

Let Y Have The Following Probability Distribution: $15 - X \quad F(x) = (15 - x)^x \cdot (x)^{15-x}, x = 0, 1, 2, \dots, 15$ Find

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Understanding the Probability Distribution of Y

When dealing with probability distributions, it's essential to understand how the random variables are defined and how their probabilities are calculated. In this case, we are given a probability distribution involving a random variable Y, which depends on another variable X. The expression " $15 - X \quad F(x) = (15 - x)^x \cdot (x)^{15-x}$ " suggests a relationship between Y and X, where the distribution of Y is derived from the distribution of X.

The goal of this article is to analyze the distribution of Y, interpret the given expression, and explore the properties and calculations associated with this distribution. We will break down the problem step-by-step, interpret the notation, and provide detailed explanations suitable for students and practitioners in probability theory.

Deciphering the Distribution Expression

The initial expression appears somewhat ambiguous: " $15 - X \quad F(x) = (15 - x)^x \cdot (x)^{15-x}$ ", $x = 0, 1, 2, \dots, 15$. To proceed, let's interpret plausible meanings:

- Possible interpretation: Y is defined as $Y = 15 - X$, where X is a discrete random variable taking values 0, 1, 2, ..., 15.
- Given: The distribution function $F(x)$ relates to X.
- Objective: Find the probability distribution of Y, i.e., $P(Y = y)$ for y in the appropriate range.

Assumption for clarity:

Suppose that X is a discrete random variable with probability mass function (pmf) $p_X(x)$, and Y is defined as $Y = 15 - X$.

Defining the Distribution of X

Before analyzing Y, we need to understand the distribution of X:

- X takes integer values from 0 to 15:

$$\begin{aligned} & \backslash[\\ & x \in \{0, 1, 2, \dots, 15\} \\ & \backslash] \end{aligned}$$

- The pmf of X can be denoted as:

$$\begin{aligned} & \backslash[\\ & p_X(x) = P(X = x) \\ & \backslash] \end{aligned}$$

- Given the notation, the cumulative distribution function (CDF) of X is:

$$\begin{aligned} & \backslash[\\ & F(x) = P(X \leq x) \\ & \backslash] \end{aligned}$$

- The problem suggests a relationship involving $F(x)$ and the value 15 minus X.

Expressing Y in Terms of X

Given $Y = 15 - X$, the possible values of Y are:

$$\begin{aligned} & \backslash[\\ & y \in \{0, 1, 2, \dots, 15\} \\ & \backslash] \end{aligned}$$

since:

$$\begin{aligned} & \backslash[\\ & X \in \{0, 1, 2, \dots, 15\} \\ & \backslash] \end{aligned}$$

and:

$$\begin{aligned} & \backslash[\\ & Y = 15 - X \\ & \backslash] \end{aligned}$$

- When $X = 0$, $Y = 15$

- When $X = 1$, $Y = 14$

- When $X = 2$, $Y = 13$

- ...

- When $X = 15$, $Y = 0$

This inverse relationship indicates that the distribution of Y is directly related to that of X , but with the support reversed.

Deriving the Distribution of Y

Because Y is a function of X , the pmf of Y can be derived from the pmf of X :

$$\begin{aligned} & \backslash[\\ & P(Y = y) = P(15 - X = y) = P(X = 15 - y) \\ & \backslash] \end{aligned}$$

for $y = 0, 1, 2, \dots, 15$.

Therefore:

$$\begin{aligned} & \backslash[\\ & \boxed{ \\ & p_Y(y) = p_X(15 - y) \\ & } \\ & \backslash] \end{aligned}$$

This simple relationship allows us to compute the distribution of Y once the distribution of X is known.

Calculating Probabilities for Y

To proceed with calculations, the key is to know $p_X(x)$. Since the problem statement doesn't specify the exact probabilities for X, we can consider common distributions or assume a uniform distribution for illustrative purposes.

Case 1: Uniform Distribution of X

Suppose X is uniformly distributed over $\{0, 1, 2, \dots, 15\}$:

$$p_X(x) = \frac{1}{16}, \quad x = 0, 1, 2, \dots, 15$$

Then, the distribution of Y is:

$$p_Y(y) = p_X(15 - y) = \frac{1}{16}$$

for $y = 0, 1, 2, \dots, 15$.

Case 2: Non-Uniform Distribution

If X follows a different distribution (e.g., binomial, geometric, etc.), the pmf $p_X(x)$ can be plugged into the relationship:

$$p_Y(y) = p_X(15 - y)$$

which allows easy computation of Y's distribution once X's distribution is known.

Expected Value and Variance of Y

Understanding the characteristics of Y , such as its expected value and variance, helps in probabilistic modeling.

Expected Value of Y

$$E[Y] = \sum_{y=0}^{15} y \cdot p_Y(y)$$

Using the relationship:

$$p_Y(y) = p_X(15 - y)$$

we get:

$$E[Y] = \sum_{y=0}^{15} y \cdot p_X(15 - y)$$

Changing variables:

Let $(k = 15 - y)$, then:

$$y = 15 - k$$

As y goes from 0 to 15, k also runs from 15 to 0. Reversing the sum:

$$E[Y] = \sum_{k=0}^{15} (15 - k) \cdot p_X(k) = 15 \sum_{k=0}^{15} p_X(k) - \sum_{k=0}^{15} k \cdot p_X(k)$$

Since the sum over all probabilities is 1:

$$\sum_{k=0}^{15} p_X(k) = 1$$

and:

$$E[X] = \sum_{k=0}^{15} k \cdot p_X(k)$$

Thus:

$$E[Y] = 15 - E[X]$$

Key insight: The expected value of Y is simply 15 minus the expected value of X.

Variance of Y

Variance is unaffected by shifting but is affected by reflection:

$$\text{Var}(Y) = \text{Var}(15 - X) = \text{Var}(X)$$

since subtracting a constant does not change variance.

Summary of Distribution Properties

Property	Expression	Explanation
Distribution of Y	$p_Y(y) = p_X(15 - y)$	Reflection of X's pmf
Support of Y	$y \in \{0, 1, 2, \dots, 15\}$	Same as X, reversed
Expected Value of Y	$E[Y] = 15 - E[X]$	Reflection property
Variance of Y	$\text{Var}(Y) = \text{Var}(X)$	Variance unchanged

Applications of the Distribution of Y

Understanding the distribution of Y provides insights in various fields:

- Statistics and Data Analysis: When transforming data, understanding how distribution properties change under transformations like $Y = 15 - X$ is essential.
- Quality Control: In manufacturing, if X represents defect counts, Y might represent remaining defect capacity.
- Game Theory and Probability Models: When modeling scenarios where outcomes are inversely related, this reflection property simplifies calculations.
- Simulation and Monte Carlo Methods: Knowing the relationship helps generate samples of Y from known samples of X efficiently.

Conclusion and Final Remarks

In summary, the probability distribution of a variable Y defined as $(Y = 15 - X)$, where X is a discrete variable over 0 to 15, can be directly derived from the distribution of X . The key takeaway is that:

- The pmf of Y is a reflected version of the pmf of X .
- The expected value of Y is 15 minus the expected value of X .
- The variance of Y equals the variance of X .

This reflection property simplifies many calculations and provides a clear pathway for analyzing related probabilistic models.

Note: The specific probabilities depend on the actual distribution of X . If the distribution of X is given explicitly, all properties and calculations can be carried out precisely. Otherwise, assuming uniform distribution provides a straightforward illustration.

Further Reading and Resources

- Probability and Statistics Textbooks: To deepen understanding of discrete distributions and transformations.
- Online Probability Calculators: For quick computation of pmfs, expected values, and variances.

- Statistical Software (e.g., R, Python): To simulate distributions and visualize the effects of transformations like $Y = 15 - X$.

By mastering the relationship between X and Y , statisticians and data analysts can efficiently analyze a wide range of scenarios involving symmetric or reflected distributions.

Frequently Asked Questions

What is the probability distribution function for the random variable Y as given?

The probability distribution function is $F(x) = (15 - x)/15$ for $x = 0, 1, 2, \dots, 15$.

How do you verify that $F(x)$ is a valid cumulative distribution function (CDF)?

To verify, check that $F(0) = 0$, $F(15) = 1$, and that $F(x)$ is non-decreasing and right-continuous within the range.

What is the probability that Y takes a specific value x in this distribution?

The probability that $Y = x$ is $P(Y = x) = F(x) - F(x - 1) = [(15 - x)/15] - [(15 - (x - 1))/15] = 1/15$ for $x = 1, 2, \dots, 15$, and $P(Y=0) = F(0) = 0$.

What is the expected value $E[Y]$ for this distribution?

$E[Y] = \text{sum of } x P(Y = x) \text{ over all } x$. Since $P(Y = x) = 1/15$ for $x = 1$ to 15 , $E[Y] = (1+2+\dots+15)/15 = (1516/2)/15 = 8$.

What is the variance $\text{Var}(Y)$ of this distribution?

$\text{Var}(Y) = E[Y^2] - (E[Y])^2$. Calculating $E[Y^2] = (1^2+2^2+\dots+15^2)/15 = (1+4+9+\dots+225)/15$. Sum of squares from 1 to 15 is $(151631)/6 = 1240$, so $E[Y^2] = 1240/15 \approx 82.67$. Therefore, $\text{Var}(Y) \approx 82.67 - 8^2 = 82.67 - 64 = 18.67$.

Is this distribution uniform or non-uniform, and why?

This distribution is uniform over the values 1 to 15 because each value has an equal probability of $1/15$, derived from the differences in the CDF.

How can this distribution be interpreted in a real-world context?

It can model scenarios where each outcome between 1 and 15 is equally likely, such as equally likely roll outcomes or evenly distributed random events within that range.

Can you find the median of this distribution?

Yes. Since the probabilities are uniform, the median is the middle value, which is $(15 + 1)/2 = 8$. Thus, the median Y is 8.

Additional Resources

Understanding the Probability Distribution of Let Y Have The Following Probability Distribution: $15 - X$,
 $F(x) = (15 - x) / (for\ x = 0, 1, 2, ..., 15)$

When working with probability distributions, especially those involving transformations of random variables, it is crucial to understand how the distribution of one variable impacts the distribution of another. In this article, we will explore a specific probability distribution characterized by the function $F(x) = (15 - x) / (for\ x = 0, 1, 2, ..., 15)$, focusing on the variable Y , which is related to X through the transformation $Y = 15 - X$. This detailed guide aims to clarify the nature of this distribution, how to derive its properties, and practical applications in statistical analysis.

Understanding the Basic Setup

Before delving into the specifics, let's clarify the notation and assumptions:

- The variable X takes integer values from 0 to 15, inclusive.
- The probability distribution function (PDF), or more precisely, the probability mass function (PMF) since X is discrete, is given as:

$$\begin{aligned} & \backslash \\ P(X = x) &= F(x) = \frac{15 - x}{\} \\ & \backslash \end{aligned}$$

where is a normalization constant to ensure the total probability sums to 1.

- The variable Y is defined as $Y = 15 - X$.

Step 1: Clarifying the Distribution of X

Determining the PMF of X

Given the form of $F(x)$, our first goal is to find the explicit formula for $P(X = x)$.

Since the distribution is over $(x = 0, 1, 2, \dots, 15)$, and the probabilities must sum to 1, we have:

$$\sum_{x=0}^{15} P(X = x) = 1$$

Given:

$$P(X = x) = \frac{1}{16} (15 - x + 1)$$

We need to find the normalization constant $\frac{1}{16}$ such that:

$$\sum_{x=0}^{15} \frac{1}{16} (15 - x + 1) = 1$$

which simplifies to:

$$\frac{1}{16} \sum_{x=0}^{15} (15 - x + 1) = 1$$

Calculating the sum:

$$\sum_{x=0}^{15} (15 - x + 1) = \sum_{x=0}^{15} 15 - \sum_{x=0}^{15} x = 16 \times 15 - \sum_{x=0}^{15} x$$

because there are 16 terms (from 0 to 15).

We know:

$$\sum_{x=0}^{15} x = \frac{15 \times 16}{2} = 120$$

Thus,

$$\sum_{x=0}^{15} (15 - x) = 16 \times 15 - 120 = 240 - 120 = 120$$

Therefore, the normalization constant is:

$$= 120$$

and the PMF of X becomes:

$$P(X = x) = \frac{15 - x}{120}$$

for $(x = 0, 1, 2, \dots, 15)$.

Step 2: Distribution of the Transformed Variable Y

Defining Y and Its Distribution

Since $(Y = 15 - X)$, the possible values of Y are:

$$Y \in \{15 - 0, 15 - 1, 15 - 2, \dots, 15 - 15\} = \{15, 14, 13, \dots, 0\}$$

which is simply the set $(\{0, 1, 2, \dots, 15\})$, but in reverse order.

Deriving the PMF of Y

The probability:

$$P(Y = y) = P(15 - X = y) = P(X = 15 - y)$$

Given the PMF of X , we get:

$$\begin{aligned} P(Y = y) &= P(X = 15 - y) = \frac{15 - (15 - y)}{120} = \frac{y}{120} \end{aligned}$$

for $(y = 0, 1, 2, \dots, 15)$.

Key insight: The distribution of Y is identical in form to X , but shifted, with probabilities proportional to y itself.

Step 3: Properties of the Distribution

3.1. Distribution of Y

The probability mass function (PMF) of Y is:

$$\begin{aligned} P(Y = y) &= \frac{y}{120} \end{aligned}$$

for $(y = 0, 1, 2, \dots, 15)$. It is a discrete distribution with probabilities increasing linearly with y .

3.2. Expected Value of Y

The expected value $(E[Y])$ is:

$$\begin{aligned} E[Y] &= \sum_{y=0}^{15} y \times P(Y=y) = \sum_{y=0}^{15} y \times \frac{y}{120} = \frac{1}{120} \sum_{y=0}^{15} y^2 \end{aligned}$$

Calculate:

$$\begin{aligned} \sum_{y=0}^{15} y^2 &= \frac{15 \times 16 \times 31}{6} = 1240 \end{aligned}$$

(using the formula for the sum of squares: $(\sum_{k=1}^n k^2 = \frac{n(n+1)(2n+1)}{6})$, for $(n=15)$).

Note that:

$$\begin{aligned} \end{aligned}$$

$$\sum_{y=0}^{15} y^2 = 0^2 + 1^2 + 2^2 + \dots + 15^2 = 1240$$

Thus,

$$E[Y] = \frac{1240}{120} = \frac{124}{12} = \frac{31}{3} \approx 10.33$$

3.3. Variance of Y

The variance is:

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

Calculate $E[Y^2]$:

$$E[Y^2] = \sum_{y=0}^{15} y^2 \times P(Y=y) = \frac{1}{120} \sum_{y=0}^{15} y^3$$

Sum of cubes:

$$\sum_{y=0}^{15} y^3 = \left(\frac{15 \times 16}{2} \right)^2 = (120)^2 = 14,400$$

since:

$$\sum_{k=1}^n k^3 = \left(\frac{n(n+1)}{2} \right)^2$$

then,

$$E[Y^2] = \frac{14,400}{120} = 120$$

Now, compute variance:

$$\begin{aligned} \text{Var}(Y) &= 120 - \left(\frac{31}{3}\right)^2 = 120 - \frac{961}{9} = \frac{1080}{9} - \frac{961}{9} = \\ &= \frac{119}{9} \approx 13.22 \end{aligned}$$

Step 4: Practical Applications and Interpretations

Understanding the distribution of Y derived from X allows for various practical interpretations:

- **Modeling Reverse Relationships:** If X represents a quantity decreasing with some process, Y can represent the residual or remaining part, which may grow linearly.
- **Simulating Distributions:** Knowing the explicit form of the PMF allows for easy simulation of Y in computational models.
- **Statistical Analysis:** Calculating moments (mean, variance) aids in hypothesis testing or parameter estimation.

Summary of Key Results

Parameter	Expression / Value	Notes
Distribution of X	$P(X=x) = \frac{15-x}{120}$	for $x=0,1,\dots,15$
Distribution of Y	$P(Y=y) = \frac{y}{120}$	for $y=0,1,\dots,15$
Expected value of Y	$E[Y] = \frac{31}{3} \approx 10.33$	
Variance of Y	$\text{Var}(Y) = \frac{119}{9} \approx 13.22$	

Final Thoughts

This analysis demonstrates how a simple linear probability distribution over a finite set can be transformed and understood through variable transformations. The distribution of $Y = 15 - X$ inherits a linear PMF, emphasizing the importance of understanding the underlying structure when analyzing probabilistic models.

By carefully deriving the distribution properties, one can apply these insights to real-world scenarios such as modeling residuals, reverse processes, or inverses in discrete probabilistic systems. The clarity in the distribution's behavior facilitates better decision-making, simulation, and statistical inference in applied contexts.

Note: The placeholder "" in the initial problem statement appears to be a typographical

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