STA2001 Probability and Statistics (I)

Lecture 9

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Review

Negative binomial distribution with parameter p and r:
X, the number of Bernoulli trials at which the rth success is observed, and its pmf takes the form of

pmf:
$$f(x) = {x-1 \choose r-1} p^r (1-p)^{x-r}, x \in \overline{S} = \{r, r+1, \cdots\}$$

▶ Poisson distribution with parameter $\lambda > 0$:

X, the number of occurrences of an event in a unit interval and its pmf takes the form of

pmf:
$$f(x) = \frac{\lambda^x e^{-\lambda}}{x!}, x \in \overline{S} = \{0, 1, \dots\}$$

Chapter 3 Continuous Distribution

Section 3.1 Random Variable of Continuous Type

Continuous RV

Recall that a RV $X:S\to \overline{S}$ is called a discrete RV if \overline{S} contains finite or countably infinite number of outcomes.

Now we consider RVs with \overline{S} that is an interval or unions of intervals, which are quite common (e.g., velocity of a vehicle traveling along the high way)

Discrete RV vs. Continuous RV

RV
$$X$$
 is a function $X: S \to \overline{S} \subseteq R$

Discrete RV:

Continuous RV:

pmf
$$f(x): \overline{S} \to (0, 1]$$

- 1. f(x) > 0
- $2. \sum_{x \in \overline{S}} f(x) = 1$
- 3. $P(X \in A) = \sum_{x \in A} f(x)$

Continuous RV

Definition

A RV X with \overline{S} that is an interval or unions of intervals is said to be continuous RV, if there exists a function $f(x):\overline{S} \to (0,\infty)$ such that

- 1. f(x) > 0, $x \in \overline{S}$
- $2. \int_{\overline{S}} f(x) dx = 1$
- 3. If $[a, b] \subseteq \overline{S}$

$$P(a \le X \le b) \stackrel{\Delta}{=} \int_a^b f(x) dx$$

f is the so called probability density function (pdf).

Discrete RV vs. Continuous RV

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Discrete RV:

pmf
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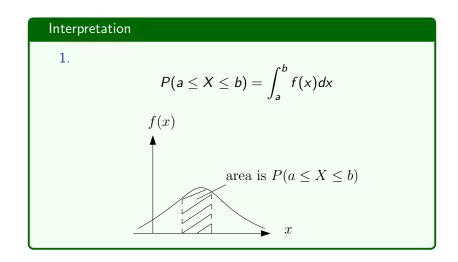
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Continuous RV:

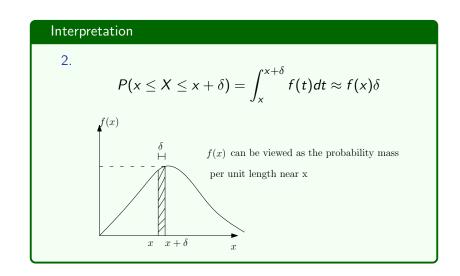
pdf
$$f(x): \overline{S} \to (0, \infty)$$

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- $2. \int_{\overline{S}} f(x) dx = 1$
- 3. $P(X \in A) = \int_A f(x) dx$

Interpretation of pdf



Interpretation of pdf



1. We often extend the domain of f(x) from \overline{S} to R and let $f(x)=0, x\notin \overline{S}$. In this case, $f(x):R\to [0,\infty)$ and \overline{S} is called the support of X.

1. We often extend the domain of f(x) from \overline{S} to R and let

$$f(x)=0, x\notin \overline{S}$$
. In this case, $f(x):R\to [0,\infty)$ and \overline{S} is called the support of X .

$$\begin{cases} f(x) \ge 0, & x \in R \\ \int_{-\infty}^{\infty} f(x) dx = 1 \\ P(a \le X \le b) = \int_{a}^{b} f(x) dx \end{cases}$$

2. For any single value a, $P(X = a) = \int_a^a f(x) dx = 0$.

Therefore, including or excluding the end points of an interval has no effect on its probability:

$$P(a \le X \le b) = P(a < X \le b) = P(a \le X < b) = P(a < X < b)$$

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3. pdf needs not to be continuous

$$f(x) = \begin{cases} \frac{1}{2}, & 0 < x < 1, & 2 < x \le 3\\ 0, & \text{otherwise} \end{cases}$$

4. pdf needs not to be bounded, e.g., the Gamma distribution

Cumulative distribution function

Definition

 $\operatorname{cdf} F(x): R \to [0,1]$

$$F(x) = P(X \le x) = \int_{-\infty}^{x} f(t)dt$$

- 1. F(x) is nondecreasing
- 2. relation between the probability function and the cdf

$$P(a \le X \le b) = F(b) - F(a)$$

relation between the pdf and the cdf

$$f(x) = F'(x)$$

for those values of x at which F(x) is differentiable $(x) = (x + 1)^{-1} + (x + 1)^{-1} +$



Example 1 [Uniform Distribution]

Let the RV X denote the outcome when a point is selected randomly from [a,b] with $-\infty < a < b < \infty$.

Define the pdf of X

$$f(x) = \begin{cases} \frac{1}{b-a}, & a \le x \le b \\ 0, & \text{otherwise} \end{cases}$$

What is the cdf of X?

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What is the cdf of X?

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \le x \le b \\ 1, & x > b \end{cases}$$

Uniform Distribution

For any
$$x \in [a, b]$$
, $P(X \le x) = \frac{x - a}{b - a}$

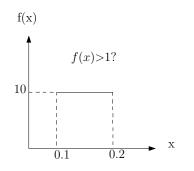
implies the probability of selecting a point from the interval [a, x] is proportional to the length of [a, x]. Such distribution is called uniform distribution and denoted by $X \sim U(a, b)$.

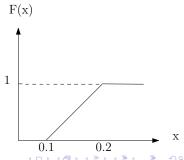
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For example, let $X \sim U(0.1, 0.2)$





Example 2, page 96

Let Y be a continuous RV with pdf g(y) = 2y, 0 < y < 1.

What is the cdf of *Y*, $P(\frac{1}{2} < Y \le \frac{3}{4})$, $P(\frac{1}{4} < Y < 2)$?

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$$g(y)$$

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$$1$$

$$1$$

$$1$$

$$P(\frac{1}{2} < Y \le \frac{3}{4}) = G(\frac{3}{4}) - G(\frac{1}{2}) = \frac{5}{16}$$

$$P(\frac{1}{4} < Y < 2) = G(2) - G(\frac{1}{4}) = \frac{15}{16}$$