Probability and Statistics: Lecture-

Monsoon-2020

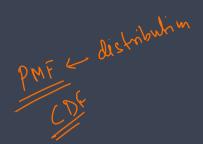
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by Pawan Kumar (IIIT, Hyderabad)
on September 11, 2020
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- 2. Special Distributions
- * Uniform Distribution
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- * Geometric Distribution
- * Binomial Distribution

Discourte



-> You may login to Skype for alternative

Attend the Quiz on Gradescope!

» Motivation for Uniform Distribution: Distribution of a Die Roll...

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Example: Motivation for Uniform Distribution

Consider rolling a fair die. The possible outcomes are $\{1, 2, 3, 4, 5, 6\}$. Then the PMF is given by

$$p(x) = \begin{cases} \frac{1}{6}, & x \in \{1, 2, 3, 4, 5, 6\} \\ 0, & otherwise \end{cases}$$

We note here that $\sum_{x \in \mathbb{Z}} |x| = 1$. We note here that PMF takes uniform values for all values of X = x.

» Uniform Distribution...

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Definition: Uniform Distribution

Motivated from the previous example, we now define uniform distribution on $\{1,2,\dots,n\}$ by

$$p(x) = \begin{cases} \frac{1}{n}, & x \in \{1, 2, \dots, n\} \\ 0, & otherwise \end{cases}$$

We verify here that $\sum_{k\in\mathbb{Z}} = 1$.



Bernoulli distribution

A random variable X is called a Bernoulli random variable with parameter p, denoted by $X \sim Bernoulli(p)$, if its PMF is given by

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- * Example: You take a pass-fail exam. You either pass or fail
- * Example: A coin is tossed, the outcome is either heads or tails

Definition of Geometric Distribution

A random variable X is called geometric random variable with parameter p, denoted by $X \sim Geometric(p)$, if its PMF is given by

$$P_X(k) = \begin{cases} p(1-p)^{k-1}, \\ 0, \end{cases}$$

for $k = 1, 2, 3, \dots$ otherwise,



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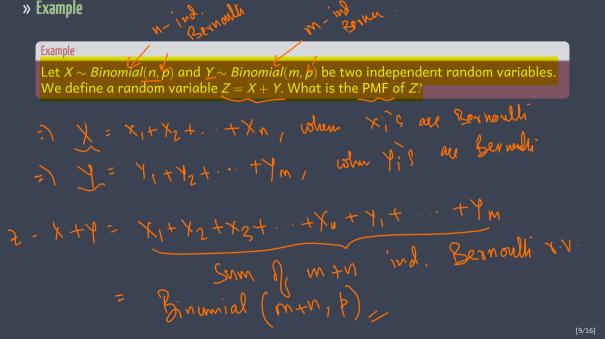
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- * We verify that $\sum_{x \in \mathbb{Z}} p(x) = \sum_{x=0}^{n} {n \choose x} p^{x} (1-p)^{n-x} = 1$



» Scratch Space...

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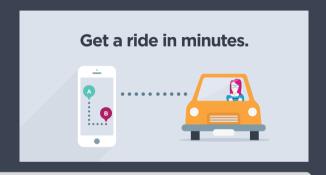
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Example

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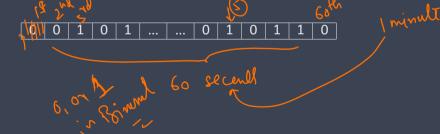
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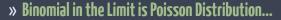
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Binomial in the Limit is Poisson Distribution...

P(
$$x=k$$
) = $\lim_{k\to\infty} \binom{n}{k} \binom{n}{n} \binom{1-k}{n} \binom{n}{k} \binom{n-k}{n} \binom$

» Definition of Poisson Distribution...

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Definition of Poisson

A random variable X is said to be a Poisson random variable with parameter λ , shown as $X \sim Poisson(\lambda)$, if its range is $R_X = \{0, 1, 2, \dots, \}$, and its PMF is given by

$$P_X(k) egin{cases} rac{e^{-\lambda}\lambda^k}{k!}, & k \in R_X \ 0 & ext{otherwise} \end{cases}$$



- * Simeon-Denis Poisson, was a French mathematician (1781-1840)
- st He published his first paper at 18, professot at 21
- * He published over 300 papers