The Bernoulli Distribution

Gene Quinn

Definition: A **Bernoulli trial** is an experiment which:

1) Has exactly two outcomes, usually called success and failure

and

2) Has a fixed probabilities p associated with the outcome "success" and q = 1 - p associated with the outcome "failure".

Definition: A **Bernoulli trial** is an experiment which:

1) Has exactly two outcomes, usually called *success* and *failure*

and

2) Has a fixed probabilities p associated with the outcome "success" and q = 1 - p associated with the outcome "failure".

The sample space for a Bernoulli trial experiment contains two elements:

 $S = \{$ success, failure $\}$

Recall that an *event* is any subset of a sample space.

In general, a finite sample space with n elements has 2^n possible subsets (Including S, each possible outcome, and the empty set \emptyset), so in the case of a Bernoulli trial there are four possible events.

Recall that a *probability function* is a real-valued function whose domain is the set of *events* associated with a sample space.

Recall that an *event* is any subset of a sample space.

In general, a finite sample space with n elements has 2^n possible subsets (Including S, each possible outcome, and the empty set \emptyset), so in the case of a Bernoulli trial there are four possible events.

Recall that a *probability function* is a real-valued function whose domain is the set of *events* associated with a sample space.

The following table lists the four possible events and defines a probability function by associating a real number with each event:

Event	Description	Probability
success ∪ failure	either success or failure occurs	1
success	the outcome is success	p
failure	the outcome is failure	q = 1 - p
Ø	neither success nor failure occurs	0

Event	Description	Probability
success ∪ failure	either success or failure occurs	1
success	the outcome is success	p
failure	the outcome is failure	q = 1 - p
Ø	neither success nor failure occurs	0

You should convince yourself that, as long as $0 \le p \le 1$, the probability function defined in tabular form above satisfies the first three Kolmogorov axioms.

Event	Description	Probability
success ∪ failure	either success or failure occurs	1
success	the outcome is success	p
failure	the outcome is failure	q = 1 - p
Ø	neither success nor failure occurs	0

You should convince yourself that, as long as $0 \le p \le 1$, the probability function defined in tabular form above satisfies the first three Kolmogorov axioms.

The fourth axiom is not required in this case because the sample space is finite.

Usually a random variable X is associated with a Bernoulli trial by the following definition.

$$X = \begin{cases} 0 & \text{if the outcome is "failure"} \\ 1 & \text{if the outcome is "success"} \end{cases}$$

Usually a random variable X is associated with a Bernoulli trial by the following definition.

$$X = \begin{cases} 0 & \text{if the outcome is "failure"} \\ 1 & \text{if the outcome is "success"} \end{cases}$$

Recall that a *random variable* is a real-valued function whose domain is a sample space, so the above definition qualifies since it is a real-valued function defined on the sample space of a Bernoulli trial,

 $S = \{$ success, failure $\}$

Finally, we will define a **probability density function** or pdf for the random variable X defined by:

$$X = \begin{cases} 0 & \text{if the outcome is "failure"} \\ 1 & \text{if the outcome is "success"} \end{cases}$$

Finally, we will define a **probability density function** or pdf for the random variable X defined by:

$$X = \begin{cases} 0 & \text{if the outcome is "failure"} \\ 1 & \text{if the outcome is "success"} \end{cases}$$

A *discrete probability density function* $p_X(k)$ maps the values of a discrete random variable X into [0, 1] according to the rule

$$p_X(k) = P(\{s \in S \mid X(s) = k\})$$

It is common to simplify the expression

$$p_X(k) = P(\{s \in S \mid X(s) = k\})$$

by removing the explicit references to the outcome \boldsymbol{s} and sample space $\boldsymbol{S}.$

The resulting simplified definition for the probability density function is:

$$p_X(k) = P(X=k)$$

It is common to simplify the expression

$$p_X(k) = P(\{s \in S \mid X(s) = k\})$$

by removing the explicit references to the outcome s and sample space S.

The resulting simplified definition for the probability density function is:

$$p_X(k) = P(X=k)$$

In the case of a Bernoulli trial, the pdf is usually defined as

$$p_X(k) = \begin{cases} p & \text{if } k = 1\\ q = 1 - p & \text{if } k = 0 \end{cases}$$

- A number of important probability distributions arise when sequences of independent Bernoulli trials with a constant probability of success p are performed.
- The most important is the result of performing exactly n Bernoulli trials, each independent of the others and having the same probability of success p.

A number of important probability distributions arise when sequences of independent Bernoulli trials with a constant probability of success p are performed.

The most important is the result of performing exactly n Bernoulli trials, each independent of the others and having the same probability of success p.

If X_i is the random variable associated with the i^{th} Bernoulli trial, for $1 \le i \le n$, then we define a **binomial random variable** Y to be

$$Y = \sum_{i=1}^{n} X_i$$

A number of important probability distributions arise when sequences of independent Bernoulli trials with a constant probability of success p are performed.

The most important is the result of performing exactly n Bernoulli trials, each independent of the others and having the same probability of success p.

If X_i is the random variable associated with the i^{th} Bernoulli trial, for $1 \le i \le n$, then we define a **binomial random variable** Y to be

$$Y = \sum_{i=1}^{n} X_i$$

Recalling that $X_i = 1$ if the *i*th Bernoulli trial results in success, we interpret *Y* as the number of successes in *n* independent Bernoulli trials with probability of success equal to *p*.

We will see that the probability density function of the binomial random variable Y obtained by adding the results X_i of n independent Bernoulli trials with probability of success p is:

$$p_Y(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, \dots, n$$

Another related distribution arises from the following experiment:

Independent Bernoulli trials each having probability of success p are performed until the first success occurs. number of important probability distributions arise when sequences of independent

A random variable *Y* is defined with the following pdf:

 $p_Y(k) = P(\text{first success occurs on trial } k), \quad k = 1, 2, 3, \dots$

The distribution of Y in this case is called the **geometric distribution**

Another related distribution arises from the following experiment:

Independent Bernoulli trials each having probability of success p are performed until the first success occurs. number of important probability distributions arise when sequences of independent

A random variable *Y* is defined with the following pdf:

 $p_Y(k) = P(\text{first success occurs on trial } k), \quad k = 1, 2, 3, \dots$

The distribution of Y in this case is called the **geometric distribution** We will see that the probability density function for the geometric distribution is:

$$p_Y(k) = (1-p)^{k-1} \cdot p, \quad k = 1, 2, \dots$$

Suppose instead we conduct independent Bernoulli trials with probability of success p until we obtain r successes, for some integer r > 1.

Define a random variable *Y* with the following pdf:

 $p_Y(k) = P$ (the r^{th} success occurs on trial k), k = r, r + 1, ...

The distribution of Y in this case is called the **negative binomial distribution**

Suppose instead we conduct independent Bernoulli trials with probability of success p until we obtain r successes, for some integer r > 1.

Define a random variable *Y* with the following pdf:

 $p_Y(k) = P(\text{the } r^{th} \text{ success occurs on trial } k), \quad k = r, r+1, \dots$

The distribution of Y in this case is called the **negative binomial distribution**

We will see that the probability density function for the negative binomial distribution is:

$$p_Y(k) = {\binom{k-1}{r-1}} p^r (1-p)^{k-r}, \quad k = r, r+1, \dots$$