Bernoulli distribution

In probability theory and statistics, the Bernoulli distribution, named after Swiss mathematician Jacob Bernoulli, [1] is the discrete probability distribution of a random variable which takes the value 1 with probability p and the value 0 with probability q = 1 - p, that is, the probability distribution of any single experiment that asks a yes-no question; the question results in a boolean-valued outcome, a single bit of information whose value is success/yes/true/one with probability p and failure/no/false/zero with probability q. It can be used to represent a (possibly biased) coin toss where 1 and 0 would represent "heads" and "tails" (or vice versa), respectively, and p would be the probability of the coin landing on heads or tails, respectively. In particular, unfair coins would have $p \neq 1/2$.

The Bernoulli distribution is a special case of the <u>binomial distribution</u> where a single trial is conducted (so *n* would be 1 for such a binomial distribution). It is also a special case of the **two-point distribution**, for which the possible outcomes need not be 0 and 1.

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Properties of the Bernoulli distribution

If X is a random variable with this distribution, then:

$$Pr(X = 1) = p = 1 - Pr(X = 0) = 1 - q.$$

The probability mass function f of this distribution, over possible outcomes k, is

$$f(k;p) = \left\{egin{array}{ll} p & ext{if } k=1,_{ ext{ iny [2]}} \ q=1-p & ext{if } k=0. \end{array}
ight.$$

This can also be expressed as

$$f(k;p) = p^k (1-p)^{1-k}$$
 for $k \in \{0,1\}$

or as

$$f(k;p) = pk + (1-p)(1-k) \quad ext{for } k \in \{0,1\}.$$

The Bernoulli distribution is a special case of the binomial distribution with $n=1.^{[3]}$

The <u>kurtosis</u> goes to infinity for high and low values of p, but for p = 1/2 the two-point distributions including the Bernoulli distribution have a lower <u>excess kurtosis</u> than any other probability distribution, namely -2.

The Bernoulli distributions for $0 \le p \le 1$ form an exponential family.

The $\underline{\mathsf{maximum}}$ likelihood estimator of $oldsymbol{p}$ based on a random sample is the $\underline{\mathsf{sample}}$ mean.

Mean

The $\underline{\mathsf{expected}}\ \mathsf{value}\ \mathsf{of}\ \mathsf{a}\ \mathsf{Bernoulli}\ \mathsf{random}\ \mathsf{variable}\ oldsymbol{X}$ is

$$E(X) = p$$

This is due to the fact that for a Bernoulli distributed random variable X with $\Pr(X=1)=p$ and $\Pr(X=0)=q$ we find

$$\mathrm{E}[X] = \Pr(X = 1) \cdot 1 + \Pr(X = 0) \cdot 0 = p \cdot 1 + q \cdot 0 = p.$$
[2]

Variance

The variance of a Bernoulli distributed $oldsymbol{X}$ is

$$\mathrm{Var}[X] = pq = p(1-p)$$

We first find

Bernoulli

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Parameters	$egin{array}{l} 0 \leq p \leq 1 \ q = 1 - p \end{array}$
Support	$k \in \{0,1\}$
pmf	$\left\{egin{array}{ll} q=1-p & ext{if } k=0 \ p & ext{if } k=1 \end{array} ight.$
CDF	$\left\{egin{array}{ll} 0 & ext{if } k < 0 \ 1-p & ext{if } 0 \leq k < 1 \ 1 & ext{if } k \geq 1 \end{array} ight.$
Mean	p
Median	$\left\{egin{array}{ll} 0 & ext{if } p < 1/2 \ [0,1] & ext{if } p = 1/2 \ 1 & ext{if } p > 1/2 \end{array} ight.$
Mode	$\left\{egin{array}{ll} 0 & ext{if } p < 1/2 \ 0,1 & ext{if } p = 1/2 \ 1 & ext{if } p > 1/2 \end{array} ight.$
Variance	p(1-p)=pq
Skewness	$rac{1-2p}{\sqrt{pq}}$
Ex. kurtosis	$\frac{1-6pq}{pq}$
Entropy	$-q \ln q - p \ln p$
MGF	$q+pe^t$
CF	$q+pe^{it}$
PGF	q + pz
Fisher information	$\frac{1}{pq}$

$$\mathrm{E}[X^2] = \Pr(X=1) \cdot 1^2 + \Pr(X=0) \cdot 0^2 = p \cdot 1^2 + q \cdot 0^2 = p$$

From this follows

$${
m Var}[X] = {
m E}[X^2] - {
m E}[X]^2 = p - p^2 = p(1-p) = pq^{[2]}$$

Skewness

The <u>skewness</u> is $\frac{q-p}{\sqrt{pq}} = \frac{1-2p}{\sqrt{pq}}$. When we take the standardized Bernoulli distributed random variable $\frac{X-\mathrm{E}[X]}{\sqrt{\mathrm{Var}[X]}}$ we find that this random variable attains $\frac{q}{\sqrt{pq}}$ with probability p and attains $-\frac{p}{\sqrt{pq}}$ with probability q. Thus we get

$$egin{aligned} \gamma_1 &= \mathrm{E} \left[\left(rac{X - \mathrm{E}[X]}{\sqrt{\mathrm{Var}[X]}}
ight)^3
ight] \ &= p \cdot \left(rac{q}{\sqrt{pq}}
ight)^3 + q \cdot \left(-rac{p}{\sqrt{pq}}
ight)^3 \ &= rac{1}{\sqrt{pq^3}} \left(pq^3 - qp^3
ight) \ &= rac{pq}{\sqrt{pq^3}} (q-p) \ &= rac{q-p}{\sqrt{pq}} \end{aligned}$$

Related distributions

• If X_1, \ldots, X_n are independent, identically distributed (<u>i.i.d.</u>) random variables, all <u>Bernoulli trials</u> with success probability p, then their <u>sum is distributed</u> according to a binomial distribution with parameters p and p:

$$\sum_{k=1}^n X_k \sim \mathrm{B}(n,p)$$
 (binomial distribution).[2]

The Bernoulli distribution is simply B(1, p), also written as Bernoulli(p).

- The categorical distribution is the generalization of the Bernoulli distribution for variables with any constant number of discrete values.
- The Beta distribution is the conjugate prior of the Bernoulli distribution.
- The geometric distribution models the number of independent and identical Bernoulli trials needed to get one success.
- If $Y \sim \operatorname{Bernoulli}\left(\frac{1}{2}\right)$, then 2Y-1 has a Rademacher distribution.

See also

- Bernoulli trials, random variables distributed according to a Bernoulli distribution
- Bernoulli process, a random process consisting of a sequence of independent Bernoulli trials
- Bernoulli sampling
- Binary entropy function
- Binomial distribution
- Binary decision diagram

References

- 1. James Victor Uspensky: *Introduction to Mathematical Probability*, McGraw-Hill, New York 1937, page 45
- 2. Bertsekas, Dimitri P. (2002). Introduction to Probability (https://www.worldcat.org/oclc/51441829). Tsitsiklis, John N., Τσιτσικλής, Γιάννης N. Belmont, Mass.: Athena Scientific. ISBN 188652940X. OCLC 51441829 (https://www.worldcat.org/oclc/51441829).
- 3. McCullagh, Peter; Nelder, John (1989). Generalized Linear Models, Second Edition. Boca Raton: Chapman and Hall/CRC. Section 4.2.2. ISBN 0-412-31760-5.

Further reading

- Johnson, N. L.; Kotz, S.; Kemp, A. (1993). *Univariate Discrete Distributions* (2nd ed.). Wiley. ISBN 0-471-54897-9.
- Peatman, John G. (1963). *Introduction to Applied Statistics*. New York: Harper & Row. pp. 162–171.

External links

- Hazewinkel, Michiel, ed. (2001) [1994], "Binomial distribution" (https://www.encyclopediaofmath.org/index.php?title=p/b016420), Encyclopedia of Mathematics, Springer Science+Business Media B.V. / Kluwer Academic Publishers, ISBN 978-1-55608-010-4
- Weisstein, Eric W. "Bernoulli Distribution" (http://mathworld.wolfram.com/BernoulliDistribution.html). *MathWorld*.
- Interactive graphic: Univariate Distribution Relationships (http://www.math.wm.edu/~leemis/chart/UDR/UDR.html)

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