

SOS POLITICAL SCIENCE AND PUBLIC ADMINISTRATION

MBA HRD 206

SUBJECT NAME: QUANTITATIVE TECHNIQUES FOR MANAGERS

UNIT-V

TOPIC NAME:

Bayes' Theorem Definition

What is the Bayes' Theorem?

Bayes' theorem, named after 18th-century British mathematician Thomas Bayes is a mathematical formula for determining conditional probability. The theorem provides a way to revise existing predictions or theories (update probabilities) given new or additional evidence. In finance, Bayes' theorem can be used to rate the risk of lending money to potential borrowers.

Bayes' theorem is also called Bayes' Rule or Bayes' Law and is the foundation of the field of Bayesian statistics.

Key Takeaways

- Bayes' Theorem allows you to update predicted probabilities of an event by incorporating new information.
- Bayes' Theorem was named after 18th century mathematician Thomas Bayes.
- It is often employed in finance in updating risk evaluation.

The Formula for Bayes' Theorem Is

$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A) \cdot P(B|A)}{P(B)}$ where: $P(A)$ = The probability of A occurring
 $P(B)$ = The probability of B occurring
 $P(A|B)$ = The probability of A given B
 $P(B|A)$ = The probability of B given A
 $P(A \cap B)$ = The probability of both A and B occurring

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 $P(A \cap B)$ = The probability of both A and B occurring
 $P(A|B) = P(B)$

$$P(A \cap B) = P(B)$$

$$P(A) \cdot P(B|A)$$

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Bayes' Theorem Explained

Applications of the theorem are widespread and not limited to the financial realm. As an example, Bayes' theorem can be used to determine the accuracy of medical test results by taking into consideration how likely any given person is to have a disease and the general accuracy of the test. Bayes' theorem relies on incorporating prior probability distributions in order to generate posterior probabilities. Prior probability, in Bayesian statistical inference, is the probability of an event before new data is collected. This is the best rational assessment of the probability of an outcome based on the current knowledge before an experiment is performed. Posterior probability is the revised probability of an event occurring after taking into consideration new information. Posterior probability is calculated by updating the prior probability by using Bayes' theorem. In statistical terms, the posterior probability is the probability of event A occurring given that event B has occurred.

Bayes' theorem thus gives the probability of an event based on new information that is, or may be related, to that event. The formula can also be used to see how the probability of an event occurring is affected by hypothetical new information, supposing the new information will turn out to be true. For instance, say a single card is drawn from a complete deck of 52 cards. The probability that the card is a king is 4 divided by 52, which equals $1/13$ or approximately 7.69%. Remember that there are 4 kings in the deck. Now, suppose it is revealed that the selected card is a face card. The probability the selected card is a king, given it is a face card, is 4 divided by 12, or approximately 33.3%, as there are 12 face cards in a deck.

Deriving Bayes' Theorem Formula With An Example

Bayes' theorem follows simply from the axioms of conditional probability. Conditional probability is the probability of an event given that another event occurred. For example, a simple probability question may ask: "What is the probability of Amazon.com, Inc., (NYSE: AMZN) stock price falling?" Conditional probability takes this question a step further by asking: "What is the probability of AMZN stock price falling given that the Dow Jones Industrial Average (DJIA) index fell earlier?"

The conditional probability of A given that B has happened can be expressed as:

If A is: "AMZN price falls" then $P(\text{AMZN})$ is the probability that AMZN falls; and B is: "DJIA is already down," and $P(\text{DJIA})$ is the probability that the DJIA fell; then the conditional probability expression reads as "the probability that

AMZN drops given a DJIA decline is equal to the probability that AMZN price declines and DJIA declines over the probability of a decrease in the DJIA index.

$$P(\text{AMZN}|\text{DJIA}) = P(\text{AMZN and DJIA}) / P(\text{DJIA})$$

$P(\text{AMZN and DJIA})$ is the probability of both A and B occurring. This is also the same as the probability of A occurring multiplied by the probability that B occurs given that A occurs, expressed as $P(\text{AMZN}) \times P(\text{DJIA}|\text{AMZN})$. The fact that these two expressions are equal leads to Bayes' theorem, which is written as:

$$\text{if, } P(\text{AMZN and DJIA}) = P(\text{AMZN}) \times P(\text{DJIA}|\text{AMZN}) = P(\text{DJIA}) \times P(\text{AMZN}|\text{DJIA})$$

$$\text{then, } P(\text{AMZN}|\text{DJIA}) = [P(\text{AMZN}) \times P(\text{DJIA}|\text{AMZN})] / P(\text{DJIA}).$$

Where $P(\text{AMZN})$ and $P(\text{DJIA})$ are the probabilities of Amazon and the Dow Jones falling, without regard to each other.

The formula explains the relationship between the probability of the hypothesis before seeing the evidence that $P(\text{AMZN})$, and the probability of the hypothesis after getting the evidence $P(\text{AMZN}|\text{DJIA})$, given a hypothesis for Amazon given evidence in the Dow.