

# Stochastic Signal Processing

## Lesson 1: Basic of Probability and Random Variables

Weize Sun

# PROBABILITY THEORY - Basics

- Probability theory deals with the study of random phenomena, which under repeated experiments yield different outcomes that have certain underlying patterns about them.
- The notion of an experiment assumes a set of repeatable conditions that allow any number of identical repetitions.
- When an experiment is performed under these conditions, certain elementary events occur in different but *completely uncertain* ways.

# PROBABILITY THEORY - Basics

- The probability of an event (事件)  $A$  is defined as:

$$P(A) \cong n_A/n \text{ } (\cong \text{ means approximate with a high degree of certainty})$$

where  $n_A$  is the number of occurrences of  $A$  and  $n$  is the total number of trials. (When  $n \rightarrow \infty$ , it is equal)

- The totality(总体) of all  $\xi_i$ , *known a priori*, constitutes a set  $S$ , the set of all experimental outcomes.

$$S = \{\xi_1, \xi_2, \dots \xi_k, \dots\}$$

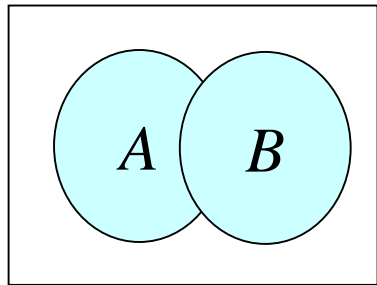
# PROBABILITY THEORY - Basics

- $S$  has subsets  $A, B, C$ . If  $A$  is a subset of  $S$ , then  $\xi \in A$  implies  $\xi \in S$ .
- And:

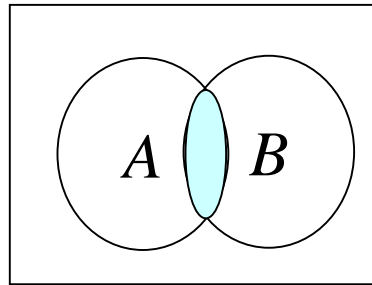
$$A \cup B = \{\xi \mid \xi \in A \text{ or } \xi \in B\}$$

$$A \cap B = \{\xi \mid \xi \in A \text{ and } \xi \in B\}$$

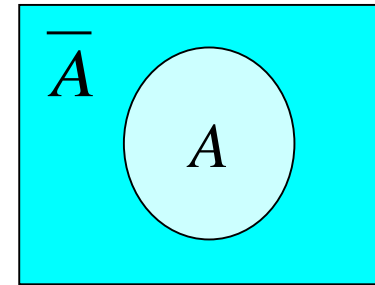
$$\overline{A} = \{\xi \mid \xi \notin A\}$$



$A \cup B$



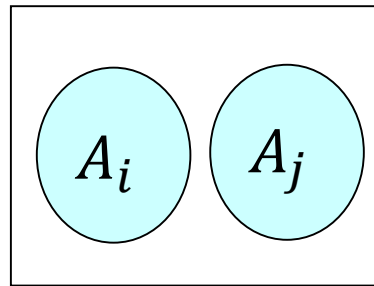
$A \cap B$



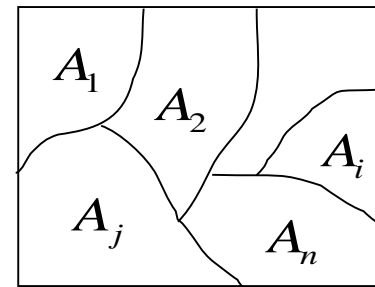
$\overline{A}$

# PROBABILITY THEORY - Basics

- If  $A \cap B = \emptyset$ , the empty set, then  $A$  and  $B$  are said to be **mutually exclusive (M.E)**.
- A partition(分割) of  $S$  is a collection of mutually exclusive subsets  $\{A_k\}$  of  $S$  such that their union(并/并集) is  $S$ .



$$A_i \cap A_j = \emptyset \text{ for } i \neq j$$



$$\bigcup_i A_i = S$$

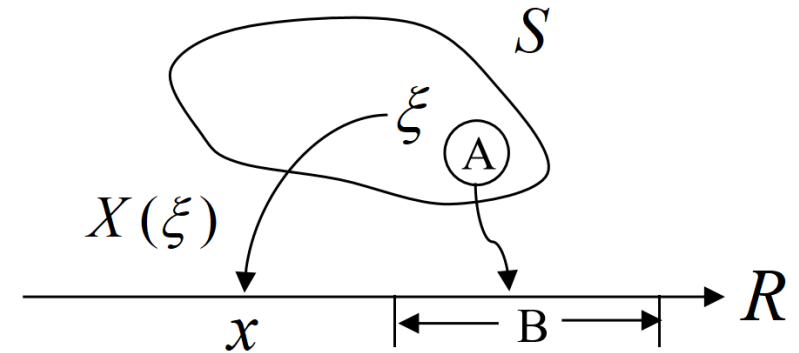
# PROBABILITY THEORY - Basics

- Suppose two subsets  $A$  and  $B$  are both events, then consider
  - “Does an outcome belongs to  $A$  or  $B$ ? ”
  - “Does an outcome belongs to  $A$  and  $B$ ? ”
  - “Does an outcome falls outside  $A$  ”?
- Thus the sets  $A \cup B, A \cap B, \overline{A}, \overline{B}$ , etc., also qualify as events.

Before we go on to probability, let's get a quick understanding of  
“what is **Stochastic**”

# Random Variables - Basics

- Let  $S$  is the set, and  $X$  is a function that maps every even  $\xi \in S$ , to a unique point  $x \in R$ , the set of real numbers.
- Since the outcome  $\xi$  is not certain, so is the value  $x \in R$ .
- If  $B$  is some subset of  $R$ , we may want to determine the probability of “ $X(\xi) \in B$ ”.
- To determine this probability, we can look at the set  $A = X^{-1}(B) \in S$  that contains all  $\xi \in S$  that maps into  $B$  under the function  $X$ .



# Random Variables - Basics

- Then: Probability of the event " $X(\xi) \in B$ " =  $P(X^{-1}(B))$ .
- The notion(概念) of random variable (r.v) makes sure that the inverse mapping  $X^{-1}(B)$  always results in an event so that we are able to determine the probability for any  $B \in \mathcal{R}$ :
- **Random Variable (r.v)**: A finite single valued function that maps the set of all experimental outcomes  $S$  into the set of real numbers  $R$  is said to be a **r.v**, if the set  $\{\xi | X(\xi) \leq x\}$  is an event for every  $x$  in  $R$ .
- Note: usually we write  $\{\xi | X(\xi) \leq x\}$  as  $\{X \leq x\}$
- This is what we discussed in 'Statistic (统计学)'



# What is Stochastic?

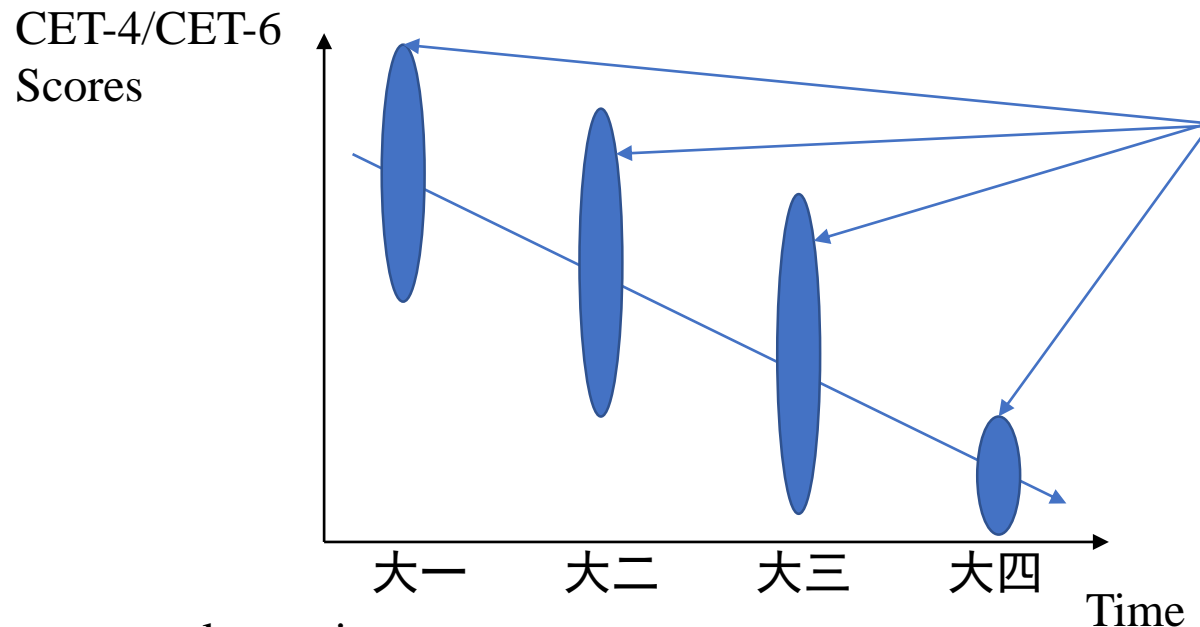
– A word from the very beginning

- Probability and random variable deal with things that are random, but its randomness will not change, for example:
  - Flipping a coin
  - Rolling a dice
  - Me taking a driver's license exam – will always fails, 100% sure, therefore it is a deterministic event (确定性事件)
  - ...

# What is Stochastic?

## – A word from the very beginning

- **Stochastic** deals with things that **change according to time and state**:
  - Your CET-4/CET-6 Scores:
  - Statistic told us that a person can get his/her highest CET-4/CET-6 Scores right after the college entrance examination (高考) \*



The probability of a person's CET-4/CET-6 Scores. For example, in Year 1, he might get scores 710 – 510, but in Year 4, he might get only 450-300.

**change according to time**

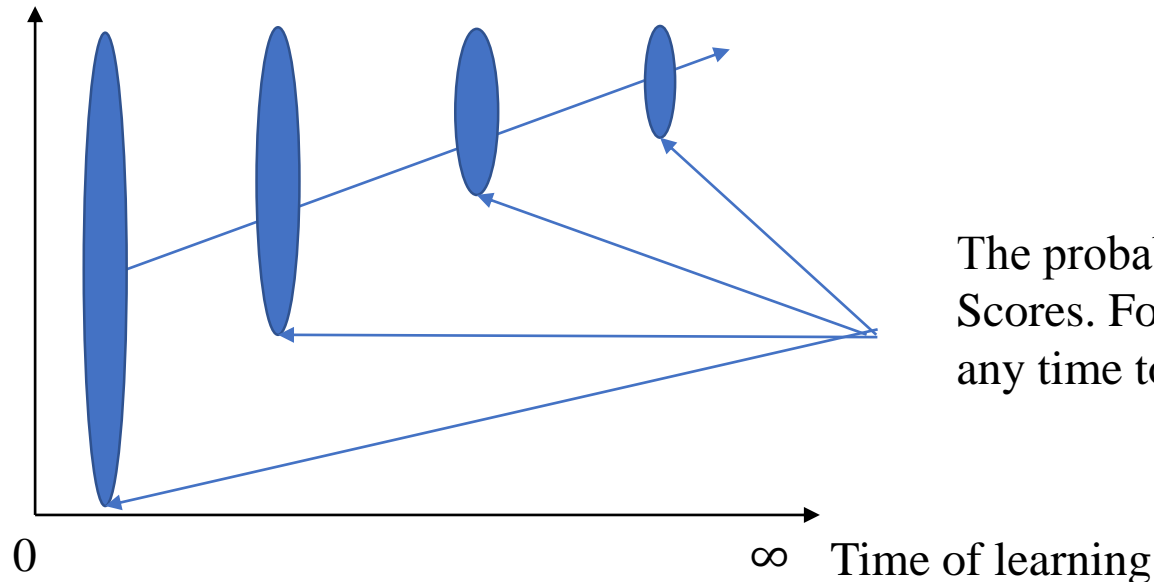
\*: just personal experience

# What is Stochastic?

– A word from the very beginning

- **Stochastic** deals with things that **change according to time and state**:
  - Your CET-4/CET-6 Scores:
  - Statistic told us that a person can get higher CET-4/CET-6 Scores under hard learning (comparing to himself only)

CET-4/CET-6  
Scores



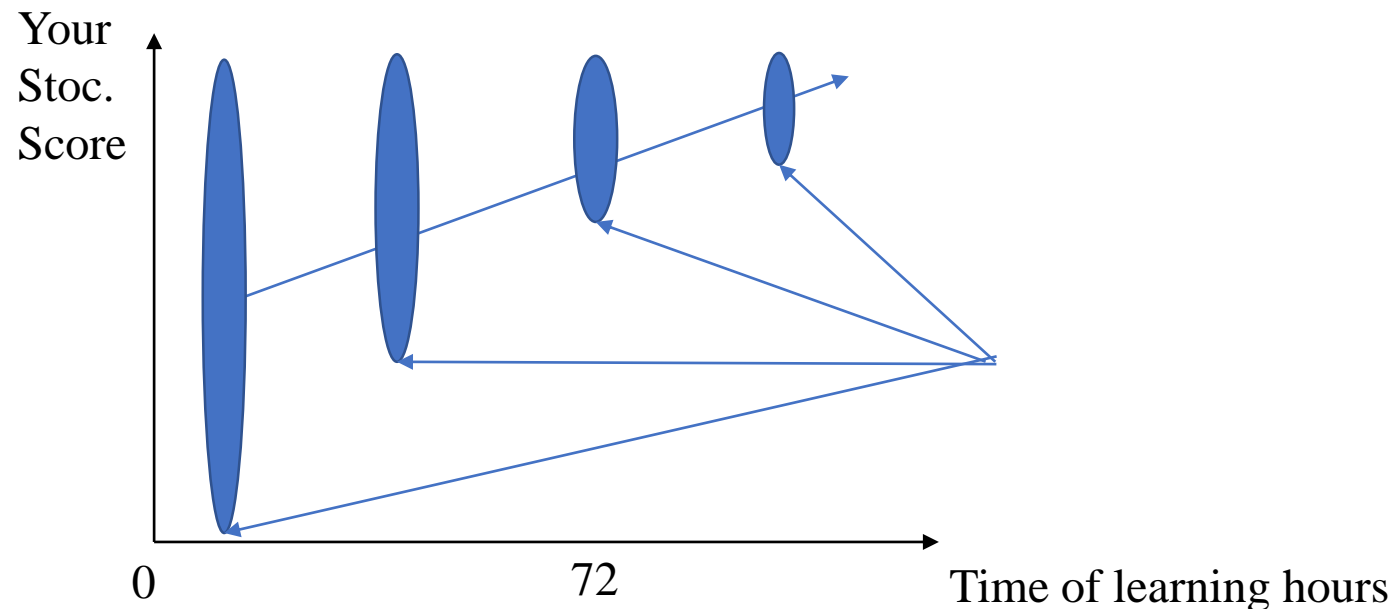
**change according to state**

The probability of a person's CET-4/CET-6 Scores. For example, if a person do not pay any time to learn it, he might get 710-300.

# What is Stochastic?

– A word from the very beginning

- Listen carefully, finish homework and experiments by yourself
- And will get a high score!

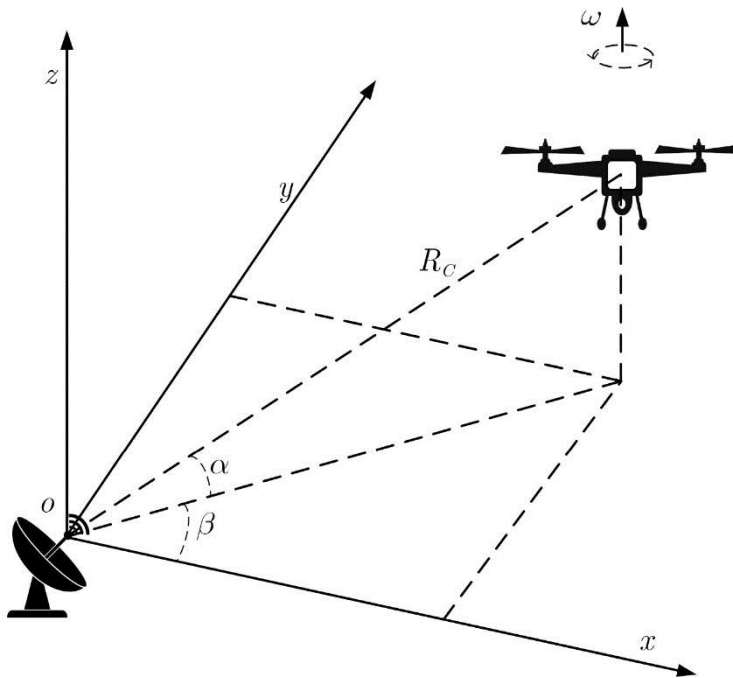


概率分析概率分，  
混沌理论混沌里；  
随机过程随机过，  
量子力学量力学；  
实变函数学十遍，  
泛函分析心犯寒；  
微机原理闹危机，  
汇编语言不会编；  
.....

# What is Stochastic?

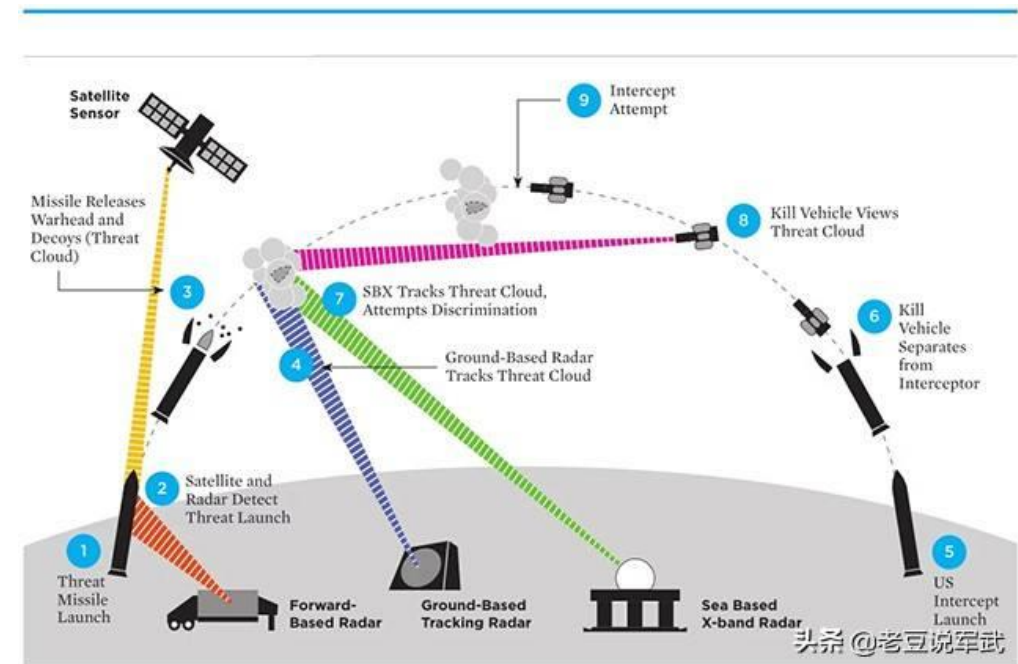
– A word from the very beginning

- **Stochastic** deals with things that **change according to time and state**
- **Stochastic** is the tool to analyze such ‘**time or state varying probability**’
  - Correlation(相关性) and Power spectrum(功率谱): mainly focus on **time varying**



Radar:

- sent out signal
- Receive the echo signal (回波信号)
- Use Time-Frequency analysis to find where the objects are
- Will try a small case of this in later experiment (experimental report required)



# What is Stochastic?

– A word from the very beginning

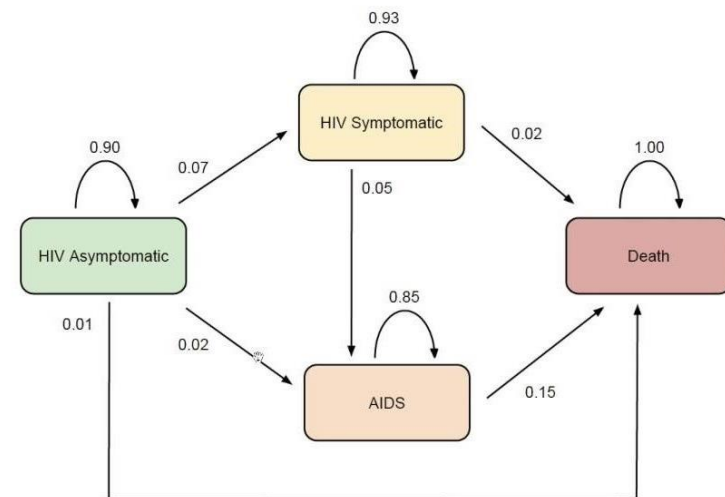
- **Stochastic** deals with things that **change according to time and state**
- **Stochastic** is the tool to analyze such ‘**time or state varying probability**’
  - Markov: mainly focus on **state** varying
  - Example: HIV expectation
  - HIV has these states:
    - HIV asymptomatic (无临床症状)
    - HIV symptomatic (有临床病状)
    - AIDS (获得性免疫缺陷综合征)
    - Death

Now we go back to probability

This stochastic matrix:

$$\begin{pmatrix} 0.90 & 0.07 & 0.02 & 0.01 \\ 0 & 0.93 & 0.05 & 0.02 \\ 0 & 0 & 0.85 & 0.15 \\ 0 & 0 & 0 & 1.00 \end{pmatrix}$$

corresponds to:



# Axioms(公理) of Probability

- For any event  $A$ , we assign a number  $P(A)$ , called the **probability** of the event  $A$ . This number satisfies the following three conditions that act the axioms of probability.
  - I.  $P(A) \geq 0$  (Probability is a nonnegative number)
  - II.  $P(S) = 1$  (Probability of the whole set is unity)
  - III. If  $A \cap B = \phi$ , then  $P(A \cup B) = P(A) + P(B)$  (1-1)

(Note that (III) states that if  $A$  and  $B$  are mutually exclusive (互不相容, M.E.) events, the probability of their union is the sum of their probabilities.)

# Conditional Probability and Independence

- In  $N$  independent trials, suppose  $N_A$ ,  $N_B$ ,  $N_{AB}$  denote the number of times events  $A$ ,  $B$  and  $AB$  occur respectively. According to the frequency interpretation of probability, for large  $N$

$$P(A) \approx \frac{N_A}{N}, \quad P(B) \approx \frac{N_B}{N}, \quad P(AB) \approx \frac{N_{AB}}{N}.$$

- Among the  $N_A$  occurrences of  $A$ , only  $N_{AB}$  of them are also found among the  $N_B$  occurrences of  $B$ . Thus the ratio

$$\frac{N_{AB}}{N_B} = \frac{N_{AB} / N}{N_B / N} = \frac{P(AB)}{P(B)}$$

is a measure of “the event  $A$  given that  $B$  has already occurred”.



# Conditional Probability and Independence

$$\frac{N_{AB}}{N_B} = \frac{N_{AB} / N}{N_B / N} = \frac{P(AB)}{P(B)}$$

- is a measure of “the event  $A$  given that  $B$  has already occurred”. We denote this conditional probability by

$P(A|B)$  = Probability of “the event  $A$  given that  $B$  has occurred”.

- We define

$$P(A|B) = \frac{P(AB)}{P(B)}, \quad (1-2)$$

- provided  $P(B) \neq 0$ . Note that the above definition satisfies all probability axioms discussed earlier.

We have

$$(i) \quad P(A|B) = \frac{P(AB)}{P(B)} \geq 0,$$

$$(ii) \quad P(S|B) = \frac{P(SB)}{P(B)} = 1, \text{ since } S \cap B = B.$$

(iii) Suppose  $A \cap C = \phi$ , Then

$$P(A \cup C | B) = \frac{P((A \cup C) \cap B)}{P(B)} = \frac{P(AB \cup CB)}{P(B)}.$$

But  $AB \cap CB = \phi$ , hence  $P(AB \cup CB) = P(AB) + P(CB)$ .

$$P(A \cup C | B) = \frac{P(AB)}{P(B)} + \frac{P(CB)}{P(B)} = P(A|B) + P(C|B),$$

$$\text{III. If } A \cap B = \phi, \text{ then } P(A \cup B) = P(A) + P(B) \quad (1-1)$$

$$P(A|B) = \frac{P(AB)}{P(B)}, \quad (1-2)$$



# Properties of Conditional Probability:

a. If  $B \subset A$ ,  $AB = B$ , and

$$P(A | B) = \frac{P(AB)}{P(B)} = \frac{P(B)}{P(B)} = 1$$

since if  $B \subset A$ , then occurrence of  $B$  implies automatic occurrence of the event  $A$ . As an example

$$A = \{\text{outcome is even}\}, B = \{\text{outcome is 2}\},$$

in a dice tossing experiment. Then  $B \subset A$ , and  $P(A | B) = 1$ .

Also, if  $A \subset B$ ,  $AB = A$ ,

# Properties of Conditional Probability:

b. If  $A \subset B$ ,  $AB = A$ , and

$$P(A | B) = \frac{P(AB)}{P(B)} = \frac{P(A)}{P(B)} > P(A).$$

(In a dice experiment,  $A = \{\text{outcome is 2}\}$ ,  $B = \{\text{outcome is even}\}$ , so that  $A \subset B$ . The statement that  $B$  has occurred (outcome is even) makes the odds for “outcome is 2” greater than without that information).

# Properties of Conditional Probability:

c. We can use the conditional probability to express the probability of a complicated event in terms of “simpler” related events.

Let  $A_1, A_2, \dots, A_n$  are pair wise disjoint and their union is  $S$ . Thus  $A_i A_j = \phi$ , and

$$\bigcup_{i=1}^n A_i = S$$

Thus  $B = B(A_1 \cup A_2 \cup \dots \cup A_n) = BA_1 \cup BA_2 \cup \dots \cup BA_n$ .

# Properties of Conditional Probability:

$$B = B(A_1 \cup A_2 \cup \dots \cup A_n) = BA_1 \cup BA_2 \cup \dots \cup BA_n.$$

- But  $A_i \cap A_j = \phi \Rightarrow BA_i \cap BA_j = \phi$ , so that:

$$P(B) = \sum_{i=1}^n P(BA_i) = \sum_{i=1}^n P(B | A_i)P(A_i). \quad (\text{全概率公式})$$

- With the notion of conditional probability, next we introduce the notion of “independence” of events.

- **Independence:**  $A$  and  $B$  are said to be independent events, if

$$P(AB) = P(A) \cdot P(B).$$

- Notice that the above definition is a probabilistic statement, *not* a set theoretic notion (集合论的概念) such as mutually exclusiveness.

# Properties of Conditional Probability:

- Suppose  $A$  and  $B$  are independent, then

$$P(A | B) = \frac{P(AB)}{P(B)} = \frac{P(A)P(B)}{P(B)} = P(A).$$

- Thus if  $A$  and  $B$  are independent, the event that  $B$  has occurred cannot provide extra information whether the event  $A$  occurred, or say, **it makes no difference to  $A$  whether  $B$  has occurred or not.** An example will clarify the situation:

# Properties of Conditional Probability:

Example 1: A box contains 6 white and 4 black balls. Remove two balls at random without replacement. What is the probability that the first one is white and the second one is black?

Let  $W_1$  = “first ball removed is white”

$B_2$  = “second ball removed is black”



# Properties of Conditional Probability:

Example 1: A box contains 6 white and 4 black balls. Remove two balls at random without replacement. What is the probability that the first one is white and the second one is black?

Let  $W_1$  = “first ball removed is white”

$B_2$  = “second ball removed is black”

- We need  $P(W_1 \cap B_2) = ?$  We have  $W_1 \cap B_2 = W_1 B_2 = B_2 W_1$ . Using the conditional probability rule,

$$P(W_1 B_2) = P(B_2 W_1) = P(B_2 | W_1) P(W_1).$$

$$P(W_1) = \frac{6}{6+4} = \frac{6}{10} = \frac{3}{5}, \quad P(B_2 | W_1) = \frac{4}{5+4} = \frac{4}{9}, \quad \Rightarrow \quad P(W_1 B_2) = \frac{3}{5} \cdot \frac{4}{9} = \frac{12}{45}$$

**Properties of Conditional Probability:**  $P(B) = \sum_{i=1}^n P(BA_i) = \sum_{i=1}^n P(B | A_i)P(A_i).$

Are the events  $W_1$  and  $B_2$  independent?



Our common sense says No.

**Properties of Conditional Probability:**  $P(B) = \sum_{i=1}^n P(BA_i) = \sum_{i=1}^n P(B | A_i)P(A_i).$

Are the events  $W_1$  and  $B_2$  independent?

To verify this we need to compute  $P(B_2)$ . Of course the fate of the second ball very much depends on that of the first ball. The first ball has two options:  $W_1$  = “first ball is white” or  $B_1$  = “first ball is black”. Note that  $W_1 \cap B_1 = \emptyset$  and  $W_1 \cup B_1 = S$ . Hence  **$W_1$  together with  $B_1$  form a partition.** Thus:

$$\begin{aligned} P(B_2) &= P(B_2 | W_1)P(W_1) + P(B_2 | B_1)P(B_1) \\ &= \frac{4}{5+4} \cdot \frac{3}{5} + \frac{3}{6+3} \cdot \frac{4}{10} = \frac{4}{9} \cdot \frac{3}{5} + \frac{1}{3} \cdot \frac{2}{5} = \frac{4+2}{15} = \frac{2}{5}, \end{aligned}$$

  $P(B_2)P(W_1) = \frac{2}{5} \cdot \frac{3}{5} \neq P(W_1B_2) = \frac{12}{45}$   **events  $W_1$  and  $B_2$  are not dependent**

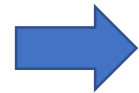
# Bayes' theorem

PPT page 18:

$$P(A | B) = \frac{P(AB)}{P(B)},$$

$$P(AB) = P(A | B)P(B).$$

$$P(B | A) = \frac{P(BA)}{P(A)} = \frac{P(AB)}{P(A)},$$



$$P(AB) = P(B | A)P(A).$$

or

$$P(A | B)P(B) = P(B | A)P(A).$$

known as **Bayes' theorem**:

$$P(A | B) = \frac{P(B | A)}{P(B)} \cdot P(A)$$

provided  $P(B) \neq 0$ . As we show below, the above definition satisfies all probability axioms discussed earlier.

# Bayes' theorem

$$P(A | B) = \frac{P(B | A) \cdot P(A)}{P(B)}$$

- Although simple enough, Bayes' theorem has an interesting interpretation:
  - $P(A)$  represents the a-priori (先验) probability of the event  $A$ . Suppose  $B$  has occurred, and assume that  $A$  and  $B$  are not independent. How can this new information be used to update our knowledge about  $A$ ?
  - Bayes' rule will take into account the new information (“ $B$  has occurred”) and gives out the a-posteriori (后验) probability of  $A$  given  $B$ .
  - We can also view the event  $B$  as new knowledge obtained from a fresh experiment. We know something about  $A$  as  $P(A)$ . The new information is available in terms of  $B$ .
  - The new information should be used to improve our knowledge/understanding of  $A$ . Bayes' theorem gives the exact mechanism for incorporating such new information.

# Bayes' theorem

A more general version of Bayes' theorem involves partition of  $S$ :

$$P(A_i | B) = \frac{P(B | A_i)P(A_i)}{P(B)} = \frac{P(B | A_i)P(A_i)}{\sum_{i=1}^n P(B | A_i)P(A_i)},$$

$A_i, i = 1 \rightarrow n$ , represent a set of mutually exclusive events with associated a-priori probabilities  $P(A_i), i = 1 \rightarrow n$ . With the new information “ $B$  has occurred”, the information about  $A_i$  can be updated by the  $n$  conditional probabilities

# Bayes' theorem

- Example 2: Two boxes  $B_1$  and  $B_2$  contain 100 and 200 light bulbs respectively. The first box ( $B_1$ ) has 15 defective bulbs and the second 5. Suppose a box is selected at random and one bulb is picked out.
  - (a) What is the probability that it is defective?
  - (b) Suppose we test the bulb and it is found to be defective. What is the probability that it came from box 1?

Try it yourself

# Bayes' theorem

- Example 2: Two boxes  $B_1$  and  $B_2$  contain 100 and 200 light bulbs respectively. The first box ( $B_1$ ) has 15 defective bulbs and the second 5. Suppose a box is selected at random and one bulb is picked out.

(a) What is the probability that it is defective?

- Solution: Note that box  $B_1$  has 85 good and 15 defective bulbs. Similarly box  $B_2$  has 195 good and 5 defective bulbs. Let  $D =$  “Defective bulb is picked out”:

$$P(D \mid B_1) = \frac{15}{100} = 0.15, \quad P(D \mid B_2) = \frac{5}{200} = 0.025.$$



# Bayes' theorem

- Since a box is selected at random, they are equally likely:

$$P(B_1) = P(B_2) = \frac{1}{2}.$$

- Thus  $B_1$  and  $B_2$  form a partition, and

$$\begin{aligned} P(D) &= P(D | B_1)P(B_1) + P(D | B_2)P(B_2) \\ &= 0.15 \times \frac{1}{2} + 0.025 \times \frac{1}{2} = 0.0875 . \end{aligned}$$

- Thus, there is 8.75% probability that a bulb picked at random is defective.

# Bayes' theorem

- (b) Suppose we test the bulb and it is found to be defective. What is the probability that it came from box 1?  $P(B_1 | D) = ?$

$$P(B_1 | D) = \frac{P(D | B_1)P(B_1)}{P(D)} = \frac{0.15 \times 1/2}{0.0875} = 0.8571 \quad .$$

- Notice that initially  $P(B_1) = 0.5$ ; then we picked out a box at random and tested a bulb that turned out to be defective. Can this information shed some light about the fact that we might have picked up box 1?
- We can see that  $P(B_1 | D) = 0.857 > 0.5$ , and indeed it is more likely at this point that we must have chosen box 1 in favor of box 2. (Recall box1 has six times more percentage of defective bulbs compared to box2).