CENG 222
Statistical Methods for Computer Engineering

Spring 2018-2019

Section 1
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Section 1 Course Web Page:
http://www.ceng.metu.edu.tr/~tcan/ceng222_s1819

Goals of the course

- Learn techniques and tools to be able to:
  - analyze and interpret large scale data,
  - apply probability theory and statistics to handle uncertainty,
  - infer facts and relationships from collected data, and
  - construct simulations by sampling from arbitrary distributions
- Acquire skills for the hot new CS field: “Data Science”

Course outline

- See the tentative schedule at:
  - http://user.ceng.metu.edu.tr/~tcan/ceng222_s1819/Schedule/index.shtml

Grading

- Midterm exam - 40%
- Final exam - 40%
- 4 Assignments (5% each) - 20%

Section 1 Course Web Site

- Syllabus
- Lecture slides and reading materials

COW

- Assignments
- Announcements at the news group: course.222
- We may also use ODTU-Class for assignments and announcements

Textbook

- Probability and Statistics for Computer Scientists, Second Edition, Michael Baron, 2013
- Your main resource of study for this course

Probability

- Studies uncertainty
- A random experiment
  - An experiment/observation which does not have a certain outcome before it is conducted
    - Examples
      - Tossing a coin
      - Observing the life time of a light bulb
      - Number of games the Cavaliers will win this season
      - Others?

Sample space

- The set of all possible outcomes of a random experiment is called the sample space
  - Tossing a coin:
    - Sample space = {H, T}
  - Tossing two coins:
    - Sample space = {HH, HT, TH, TT}
  - Lifetime of a light bulb:
    - Sample space = $[0,+\infty)$

Event

- Any collection of possible outcomes of an experiment
  - Any subset of the sample space
- Examples:
  - Experiment: tossing two coins. Event: obtaining exactly one head. $\{HT,TH\} \subset \{HH,HT,TH,TT\}$
  - Experiment: lifetime of light bulb. Event: light bulb does not last more than a month.

    $[0,1] \subset [0,+\infty)$

Event

- A sample space of N possible outcomes yields 2^N possible events
- Example: tossing a dice once
- Sample space = $\{1,2,3,4,5,6\}$
- Number of possible events = $2^6 = 64$
- Example events?

Notation used in the book

- Ω = sample space
- $\varnothing$ = empty event
- $\boldsymbol{P}\{E\}$ = probability of event E

Event algebra

- Union of two events: same as set union
  - $A \cup B = \{x\colon x \in A \text{ or } x \in B\}$
- Intersection of two events: same as set intersection
  - $A \cap B = \{x\colon x \in A \text{ and } x \in B\}$
- Complementation: same as in sets
  - A^c or $\bar{A} = \{x\colon x \in \Omega \text{ and } x \notin A\}$
- Difference: same as in sets
  - $\mathrm{A} \backslash \mathrm{B} = \{\mathrm{x}\colon x \in A \text{ and } x \notin B\}$

Disjoint and exhaustive events

- Disjoint events: If A and B have no outcomes in common, i.e., $A \cap B = \varnothing$
  – Also called mutually exclusive events
- If the union of a number of events equals the sample space, they are called exhaustive
  – $A \cup B \cup C = \Omega$

Complement, Union, Intersection

- $\overline{A \cup B} = \bar{A} \cap \bar{B}$
- $\overline{A \cap B} = \bar{A} \cup \bar{B}$
- $\overline{E_1 \cup E_2 \cup E_3 \cup E_4} = \overline{E_1} \cap \overline{E_2} \cap \overline{E_3} \cap \overline{E_4}$
- $\overline{E_1 \cap E_2 \cap E_3 \cap E_4} = \overline{E_1} \cup \overline{E_2} \cup \overline{E_3} \cup \overline{E_4}$

Probability

- Assignment of a real number to an event
  - The relative frequency of occurrence of an event in a large number of experiments
- $\boldsymbol{P}(A)$
- Axioms of probability:
  - $\mathbf{P}(A) \geq 0$
  - $\mathbf{P}(\boldsymbol{\Omega}) = 1$
  - If A and B are mutually exclusive events, then $\boldsymbol{P}(A \cup B) = \mathbf{P}(A) + \mathbf{P}(B)$
- Any function that satisfies these axioms is called a probability function

Example

- Experiment:
  - Tossing two coins
  - A = {obtaining exactly one head}
  - $\boldsymbol{P}(A) = ?$

Computing probabilities

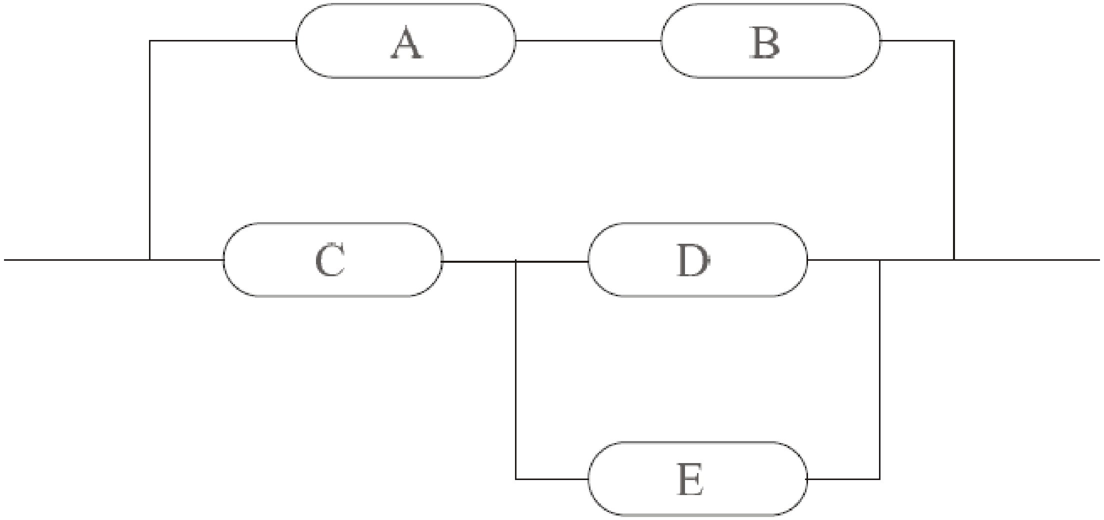
- for non-“mutually exclusive” events:
  - $\boldsymbol{P}(A \cup B) = \boldsymbol{P}(A) + \boldsymbol{P}(B) - \boldsymbol{P}(A \cap B)$

Independent Events

- $\boldsymbol{P}(E_1 \cap E_2 \cap E_3) = P\{E_1\} \cdot P\{E_2\} \cdot P\{E_3\}$

Applications in reliability

- Example 2.18
- Example 2.19
- Example 2.20



Conditional probability

- Updating of the sample space based on new information
- Consider two events A and B. Suppose that the event B has occurred. This information will change the probability of event A.
- $\boldsymbol{P}(A|B)$ denotes the conditional probability of event A given that B has occurred.

Conditional probability

- If A and B are events in Ω and $\boldsymbol{P}(B)>0$, then $\mathbf{P}(A|B)$ is called the conditional probability of A given B if the following axiom is satisfied:
  - $\boldsymbol{P}(A|B) = \boldsymbol{P}(A \cap B)/\boldsymbol{P}(B)$
- Example: tossing a fair dice.
  - A = {the number on the dice is even}
  - B = {the number on the dice < 4}
  - $\boldsymbol{P}(A|B) = ?$

Independence

- If $\boldsymbol{P}(A|B)=\boldsymbol{P}(A)$ we call that event A is independent of event B
- Note:
  - if two events A and B are independent, then $\boldsymbol{P}(A\cap B)=\boldsymbol{P}(A)\boldsymbol{P}(B)$
- Show that $\boldsymbol{P}(B|A)=\boldsymbol{P}(B)$ also holds in this case.
  - In other words, A and B are mutually independent
- This does NOT mean that they are disjoint. If A and B are disjoint then $\boldsymbol{P}(B|A)=0$

Independence

- Example: tossing a fair dice.
  - A = {the number on the dice is even}
  - B = {the number on the dice > 2}
  - $P(A|B) = ?$
  - $P(B|A) = ?$
  - $P(A) = ?$
  - $P(B) = ?$
- Example 2.31

Bayes' Rule

- Using conditional probability formula we may write:
  - $\boldsymbol{P}(A|B) = \boldsymbol{P}(A \cap B)/\boldsymbol{P}(B)$
  - $\boldsymbol{P}(B|A) = \boldsymbol{P}(A \cap B)/\boldsymbol{P}(A)$
  - $\rightarrow \boldsymbol{P}(A \cap B) = \boldsymbol{P}(A|B)\boldsymbol{P}(B) = \boldsymbol{P}(B|A)\boldsymbol{P}(A) \rightarrow$

    $\boldsymbol{P}(B|A) = \boldsymbol{P}(A|B)\boldsymbol{P}(B) \,/\, \boldsymbol{P}(A)$
- This is known as the Bayes' rule
- It forms the basis of Bayesian statistics
- What additional probabilities do we need to know to solve Example 2.32?

Law of Total Probability

- Let B_1, B_2, B_3, ….,B_k be a partition of the sample space. B_is are mutually disjoint. Let A be any event.
- Note that B_is also partition A
- Then for each $i = 1,2,\ldots,k$

$$P(B_i \mid A) = \frac{P(A \mid B_i)P(B_i)}{P(A)} = \frac{P(A \mid B_i)P(B_i)}{\sum_{j=1}^{k} P(A \mid B_j)P(B_j)}$$

When ***P***(A) is not directly known, but known conditionally, we make use of this law.

Bayes' Rule for two events

$$P(B \mid A) = \frac{P(A \mid B)P(B)}{P(A)} = \frac{P(A \mid B)P(B)}{P(A \mid B)P(B) + P(A \mid \overline{B})P(\overline{B})}$$

- Now, solve Exercise 2.32, given $\boldsymbol{P}(B)$

Another example

- A novel disease diagnostic kit is 95% effective in detecting a certain disease when it is present. The test also has a 1% false positive rate. If 0.5% of the population has the disease, what is the probability a person with a positive test result actually has the disease?

Solution

- A = {a person's test result is positive}
- B = {a person has the disease}
- $\boldsymbol{P}(B) = 0.005$, $\boldsymbol{P}(A|B) = 0.95$, $\boldsymbol{P}(A|B^c) = 0.01$

$$P(B \mid A) = \frac{P(A \mid B)P(B)}{P(A \mid B)P(B) + P(A \mid B^c)P(B^c)}$$

$$= \frac{0.95 \times 0.005}{0.95 \times 0.005 + 0.01 \times (1 - 0.005)} = \frac{475}{1470} \approx 0.323$$

Random Variables

- A random variable (r.v.) associates a unique numerical value with each outcome in the sample space. It is a real-valued function from a sample space Ω into real numbers.
- Similar to events it is denoted by an uppercase letter (e.g., X or Y) and a particular value taken by a r.v. is denoted by the corresponding lowercase letter (e.g., x or y).

Examples

- Toss three coins. X = number of heads
- Pick a student from the Computer Engineering Department.

  X = age of the student
- Observe lifetime of a light bulb

  X = lifetime in minutes
- X may be discrete or continuous