# Department of Statistics Faculty of Science Yarmouk University 

## SATS 101

## introduction to probability

and statistics

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# Chapter 4 <br> Probability and Probability Distributions 

## Basic Concepts:

- An experiment is the process by which an observation (or measurement) is obtained.
- Experiment : Record an age
: Toss a die
: Record an opinion (yes, no)
: Toss two coins
- A simple event is the outcome that is observed on a single repetition of the experiment.
- The basic element to which probability is applied.
- One and only one simple event can occur when the experiment is performed.
- A simple event is denoted by E with a subscript.
- Each simple event will be assigned a probability, measuring "how often" it occurs.
- The set of all simple events of an experiment is called the sample space, $\mathbf{S}$.
- An event is a collection of one or more simple events.
- Two events are mutually exclusive if, when one event occurs, the other cannot, and vice versa.


## The Probability of an Event

- The probability of an event A measures "how often" we think A will occur. We write $\mathbf{P}(\mathbf{A})$.

Suppose that an experiment is performed $n$ times. The relative frequency for an event A is

- If we let $n$ get infinitely large,
$P(A)=\lim _{n \rightarrow \infty} \frac{f}{n}$
- $\quad \mathrm{P}(\mathrm{A})$ must be between 0 and 1 .
- If event A can never occur, $\mathrm{P}(\mathrm{A})=0$. If event A
always occurs when the experiment is performed, $\mathrm{P}(\mathrm{A})$
$=1$.
- The sum of the probabilities for all simple events in $S$ equals 1 .
- The probability of an event $\mathbf{A}$ is found by adding the probabilities of all the simple events contained in A.


## Counting Rules

- If the simple events in an experiment are equally likely, you can calculate
$P(A)=\frac{n_{A}}{N}=\frac{\text { number of simple events in A }}{\text { total number of simple events }}$
- You can use counting rules to find $n_{A}$ and $N$.


## Example:

Toss three coins. The total number of simple events is:
$\mathbf{2} \times \mathbf{2 \times 2}=8$

## Example:

Toss two dice. The total number of simple events is: $\mathbf{6 \times 6}=\mathbf{3 6}$

## Example:

Two M\&Ms are drawn from a dish containing two red and two blue candies. The total number of simple events is: $\mathbf{4 \times 3}=\mathbf{1 2}$

## Event Relations

- The complement of an event $\mathbf{A}$ consists of all outcomes of the experiment that do not result in event $A$. We write $\mathbf{A}^{\mathrm{C}}$.



## Example

Select a student from the classroom and
record his/her hair color and gender.
Mutually exclusive; $\mathrm{B}=\mathrm{C}^{\mathrm{C}}$

> A: student has brown hair

B: student is female

- C: student is male
- What is the relationship between events $\mathbf{B}$ and $\mathbf{C}$ ?
- $\quad \mathbf{A}^{\mathrm{C}}$ : Student does not have brown hair
- $\quad B \cap C$ : Student is both male and female $=\varnothing$
- $\quad \mathbf{B} \cup \mathbf{C}$ : Student is either male and female $=$ all students $=$ S


## Calculating Probabilities for Unions and Complements

- There are special rules that will allow you to calculate probabilities for composite events.
- The Additive Rule for Unions:
- For any two events, $\mathbf{A}$ and $\mathbf{B}$, the probability of their union, $\mathbf{P}(\mathbf{A} \cup \mathbf{B})$, is $P(A \cup B)=P(A)+P(B)-P(A \cap B)$


## Example: Additive Rule

Suppose that there were 120 students in the classroom, and that they could be classified as follows:

|  | Brown | Not Brown |
| :---: | :---: | :---: |
| Male | 20 | 40 |
| Female | $\mathbf{3 0}$ | $\mathbf{3 0}$ |

A: brown hair, $\mathbf{P}(\mathbf{A})=\mathbf{5 0 / 1 2 0}$
$B$ : female, $\mathbf{P}(\mathbf{B})=\mathbf{6 0 / 1 2 0}$
Check: $\mathrm{P}(\mathrm{A} \cup \mathrm{B})$

$$
=(20+30+30) / 120
$$

$\mathrm{P}(\mathrm{A} \cup \mathrm{B})=\mathrm{P}(\mathrm{A})+\mathrm{P}(\mathrm{B})-\mathrm{P}(\mathrm{A} \cap \mathrm{B})$
$=50 / 120+60 / 120-30 / 120$
$=80 / 120=2 / 3$

## A Special Case

When two events A and B are mutually exclusive, $\mathrm{P}(\mathrm{A} \cap \mathrm{B})=0$ and $\mathbf{P}(\mathbf{A} \cup \mathbf{B})=\mathbf{P}(\mathbf{A})+\mathbf{P}(\mathbf{B})$.

|  | Brown | Not Brown |
| :---: | :---: | :---: |
| Male | 20 | 40 |
| Female | 30 | 30 |

A: male with brown hair

$$
\mathrm{P}(\mathrm{~A})=20 / 120
$$

B: female with brown hair

$$
\mathrm{P}(\mathrm{~B})=30 / 120
$$

$A$ and $B$ are mutually exclusive, so that
$\mathrm{P}(\mathrm{A} \cup \mathrm{B})=\mathrm{P}(\mathrm{A})+\mathrm{P}(\mathrm{B})$
$=20 / 120+30 / 120$
$=50 / 120$

## Calculating Probabilities for Complements

- We know that for any event $\mathbf{A}$ :

$$
-\quad \mathbf{P}\left(\mathbf{A} \cap \mathbf{A}^{\mathrm{C}}\right)=\mathbf{0}
$$

- Since either $\mathbf{A}$ or $\mathbf{A C}$ must occur, $\mathbf{P}\left(\mathbf{A} \cup \mathbf{A}^{\mathrm{C}}\right)=\mathbf{1}$
- $\quad$ so that $\mathbf{P}\left(\mathbf{A} \cup \mathbf{A}^{\mathrm{C}}\right)=\mathbf{P}(\mathbf{A})+\mathbf{P}\left(\mathbf{A}^{\mathrm{C}}\right)=\mathbf{1}$
$\mathbf{P}\left(\mathbf{A}^{\mathbf{C}}\right)=\mathbf{1}-\mathbf{P}(\mathbf{A})$


## Example

Select a student at random from the classroom. Define:

|  | Brown | Not Brown |
| :---: | :---: | :---: |
| Male | $\mathbf{2 0}$ | 40 |
| Female | $\mathbf{3 0}$ | $\mathbf{3 0}$ |

A: male

$$
\mathrm{P}(\mathrm{~A})=60 / 120
$$

B: female
$A$ and $B$ are complementary, so that $\mathrm{P}(\mathrm{B})=1-\mathrm{P}(\mathrm{A})$
$=1-60 / 120=60 / 120$

## Calculating Probabilities for Intersections

- In the previous example, we found $\mathrm{P}(\mathrm{A} \cap \mathrm{B})$ directly from the table. Sometimes this is impractical or impossible. The rule for calculating $\mathrm{P}(\mathrm{A} \cap \mathrm{B})$ depends on the idea of independent and dependent events.
Two events, $\mathbf{A}$ and $\mathbf{B}$, are said to be independent if and only if the probability that event $\mathbf{A}$ occurs does not change, depending on whether or not event $\mathbf{B}$ has occurred.


## Example 1

- Toss a fair coin twice. Define
- A: head on second toss
- B: head on first toss
$\mathrm{P}(\mathrm{A} \mid \mathrm{B})=1 / 2$
$\mathrm{P}(\mathrm{A} \mid$ not B$)=1 / 2$
$\mathrm{P}(\mathrm{A})$ does not change, whether B happens or not..
A and B are independent!


## Example 2

- A bowl contains five $M \& M s ®$, two red and three blue.

Randomly select two candies, and define

- A: second candy is red.
- B: first candy is blue.
$\mathrm{P}(\mathrm{A} \mid \mathrm{B})=\mathrm{P}(2$ nd red $\mid 1$ st blue $)=2 / 4=1 / 2$
$\mathrm{P}(\mathrm{A} \mid$ not B$)=\mathrm{P}(2$ nd red $\mid 1$ st red $)=1 / 4$
$\mathrm{P}(\mathrm{A})$ does change, depending on whether B happens or not...
A and B are dependent!


## Defining Independence

We can redefine independence in terms of conditional probabilities:
Two events $A$ and $B$ are independent if and only if

$$
\mathrm{P}(\mathrm{~A} \mid \mathrm{B})=\mathrm{P}(\mathrm{~A}) \quad \text { or } \quad \mathrm{P}(\mathrm{~B} \mid \mathrm{A})=\mathrm{P}(\mathrm{~B})
$$

Otherwise, they are dependent.

- Once you've decided whether or not two events are independent, you can use the following rule to calculate their intersection.


## The Multiplicative Rule for Intersections

- For any two events, $\mathbf{A}$ and $\mathbf{B}$, the probability that both $\mathbf{A}$ and $\mathbf{B}$ occur is
$\mathbf{P}(\mathbf{A} \cap \mathbf{B})=\mathbf{P}(\mathbf{A}) \mathbf{P}(\mathbf{B}$ given that $\mathbf{A}$ occurred $)=\mathbf{P}(\mathbf{A}) \mathbf{P}(\mathbf{B} \mid \mathbf{A})$
- If the events $\mathbf{A}$ and $\mathbf{B}$ are independent, then the probability that both $\mathbf{A}$ and $\mathbf{B}$ occur is
$\mathbf{P}(\mathbf{A} \cap \mathbf{B})=\mathbf{P}(\mathbf{A}) \mathbf{P}(\mathbf{B})$


## Example 1

In a certain population, $10 \%$ of the people can be classified as being high risk for a heart attack. Three people are randomly selected from this population. What is the probability that exactly one of the three are high risk?
Define H: high risk N : not high risk $\mathrm{P}($ exactly one high risk $)=\mathrm{P}(\mathrm{HNN})+\mathrm{P}(\mathrm{NHN})+\mathrm{P}(\mathrm{NNH})$
$=\mathrm{P}(\mathrm{H}) \mathrm{P}(\mathrm{N}) \mathrm{P}(\mathrm{N})+\mathrm{P}(\mathrm{N}) \mathrm{P}(\mathrm{H}) \mathrm{P}(\mathrm{N})+\mathrm{P}(\mathrm{N}) \mathrm{P}(\mathrm{N}) \mathrm{P}(\mathrm{H})$
$=(.1)(.9)(.9)+(.9)(.1)(.9)+(.9)(.9)(.1)=3(.1)(.9)^{2}=.243$

## Example 2

Suppose we have additional information in the previous example. We know that only $49 \%$ of the population are female. Also, of the female patients, $8 \%$ are high risk. A single person is selected at random. What is the probability that it is a high risk female?
Define H: high risk $\quad F$ : female
From the example, $\mathrm{P}(\mathrm{F})=.49$ and $\mathrm{P}(\mathrm{H} \mid \mathrm{F})=.08$. Use the
Multiplicative Rule:
$\mathrm{P}($ high risk female $)=\mathrm{P}(\mathrm{H} \cap \mathrm{F})$
$=\mathrm{P}(\mathrm{F}) \mathrm{P}(\mathrm{H} \mid \mathrm{E})=.49(.08)=.0392$

## The Mean and Standard Deviation

- Let $x$ be a discrete random variable with probability distribution $p(x)$. Then the mean, variance and standard deviation of $x$ are given as
Mean : $\mu=\Sigma x p(x)$
Variance : $\sigma^{2}=\boldsymbol{\Sigma}(x-\mu)^{2} p(x)$
Standard deviation : $\sigma=\sqrt{\sigma^{2}}$


## Example

- Toss a fair coin 3 times and record $x$ the number of heads.

| $x$ | $p(x)$ | $x p(x)$ | $(x-\mu) 2 p(x)$ |
| :--- | :--- | :--- | :--- |
| 0 | $1 / 8$ | 0 | $(-1.5) 2(1 / 8)$ |
| 1 | $3 / 8$ | $3 / 8$ | $(-0.5) 2(3 / 8)$ |
| 2 | $3 / 8$ | $6 / 8$ | $(0.5) 2(3 / 8)$ |
| 3 | $1 / 8$ | $3 / 8$ | $(1.5) 2(1 / 8)$ |

$\mu=\sum x p(x)=\frac{12}{8}=1.5$
$\sigma^{2}=\Sigma(x-\mu)^{2} p(x)$
$\sigma^{2}=.28125+.09375+.09375+.28125=.75$
$\sigma=\sqrt{.75}=.688$

## Example

- The probability distribution for $x$ the number of heads in tossing 3 fair coins.

- Shape? Symmetric; mound-shaped
- Outliers? None
- Center? $\mathrm{m}=1.5$
- Spread? s = . 688


## Key Concepts

## I. Experiments and the Sample Space

1. Experiments, events, mutually exclusive events, simple events
2. The sample space
3. Venn diagrams, tree diagrams, probability tables

## II. Probabilities

1. Relative frequency definition of probability
2. Properties of probabilities
a. Each probability lies between 0 and 1 .
b. Sum of all simple-event probabilities equals 1 .
3. $\mathrm{P}(\mathrm{A})$, the sum of the probabilities for all simple events in A

## III. Counting Rules

1. $m n$ Rule; extended $m n$ Rule
2. Permutations: $P_{r}^{n}=\frac{n!}{(n-r)!}$
3. Combinations: $C_{r}^{n}=\frac{n!}{r!(n-r)!}$

## IV. Event Relations

1. Unions and intersections
2. Events
a. Disjoint or mutually exclusive: $P(A \cap B)=0$
b. Complementary: $P(A)=1-P(A C)$
3. Conditional probability: $P(A \mid B)=\frac{P(A \cap B)}{P(B)}$
4. Independent and dependent events
5. Additive Rule of Probability:
$P(A \cup B)=P(A)+P(B)-P(A \cap B)$
6. Multiplicative Rule of Probability: $P(A \cap B)=P(A) P(B \mid A)$
7. Law of Total Probability
8. Bayes' Rule

## V. Discrete Random Variables and Probability Distributions

1. Random variables, discrete and continuous
2. Properties of probability distributions

$$
0 \leq p(x) \leq 1 \text { and } \sum p(x)=1
$$

3. Mean or expected value of a discrete random variable:

Mean : $\mu=\sum x p(x)$
4. Variance and standard deviation of a discrete random
variable:

$$
\begin{aligned}
& \text { Variance }: \sigma^{2}=\Sigma(x-\mu)^{2} p(x) \\
& \text { Standard deviation }: \sigma=\sqrt{\sigma^{2}}
\end{aligned}
$$

