

CEE 260/MIE 273: Probability and Statistics in Civil Engineering

Lecture 2b: Theory of Probability

Prof. Oke

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College of Engineering

September 16, 2025

Outline

- ① Theory of Probability
- ② Addition Rule
- ③ Counting methods
- ④ Outlook
- ⑤ Appendix

Recap from Lecture 2a

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- De Morgan's Rules are useful for expressing complements of unions or of intersections:

$$\begin{aligned}\overline{A \cup B} &= \bar{A} \cap \bar{B} \\ \overline{A \cap B} &= \bar{A} \cup \bar{B}\end{aligned}$$

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- Perform data input, set operations, permutations and combinations in MATLAB

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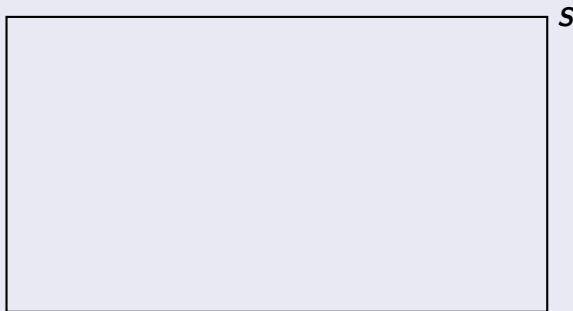
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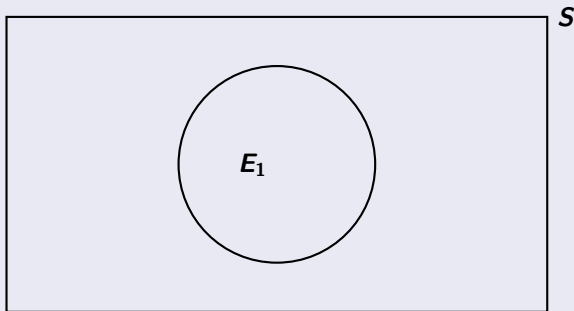
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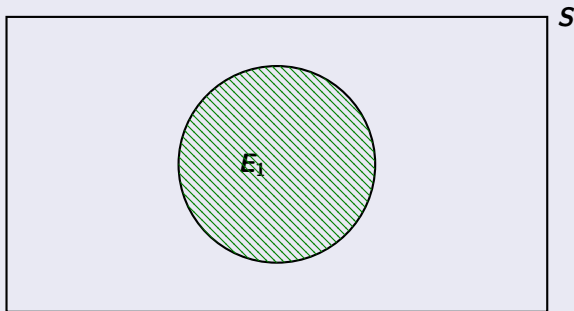
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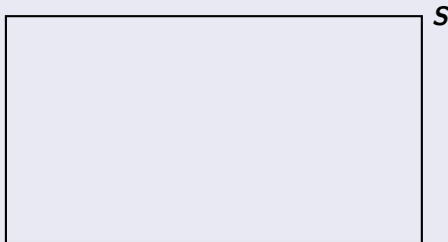
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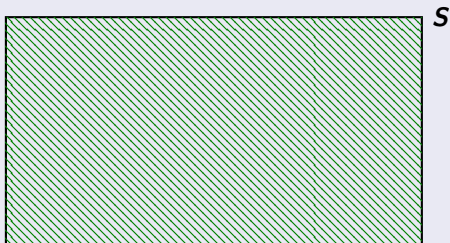
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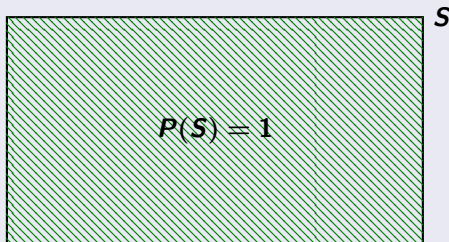
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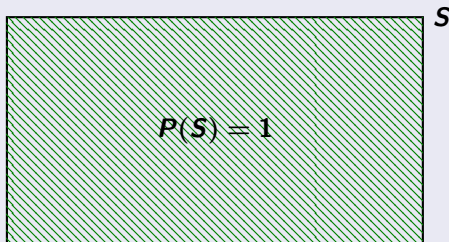
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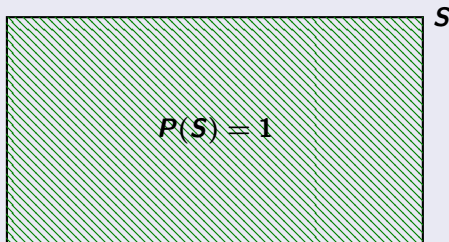
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In other words, the probability of the sample space is unity (or 1).

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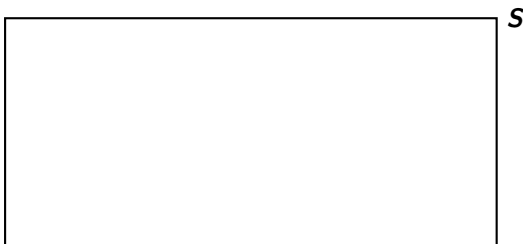
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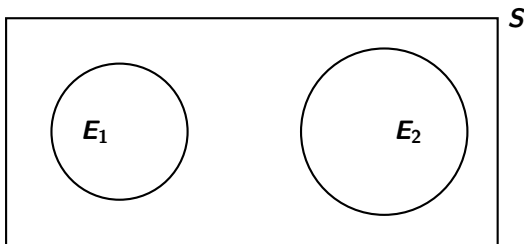
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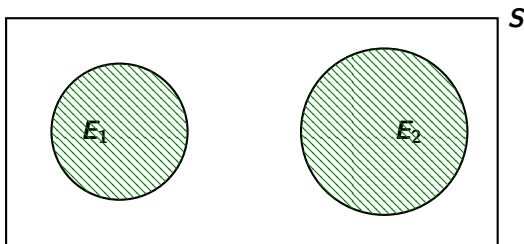
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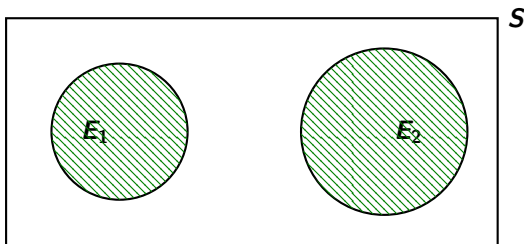
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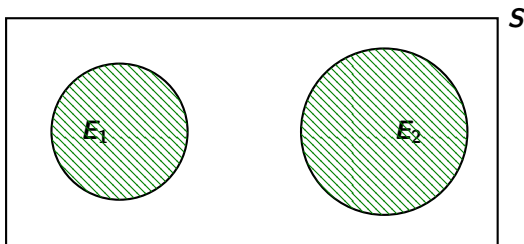
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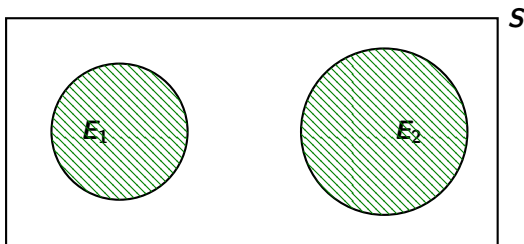
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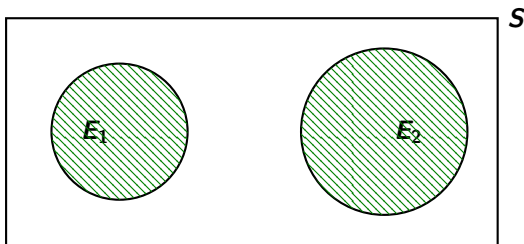
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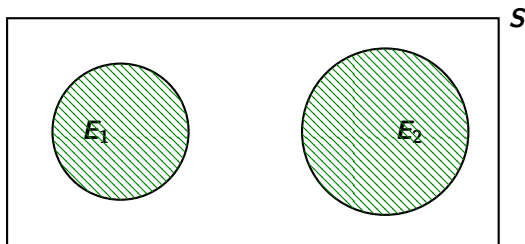
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$$0 \leq P(E) \leq 1 \quad (5)$$

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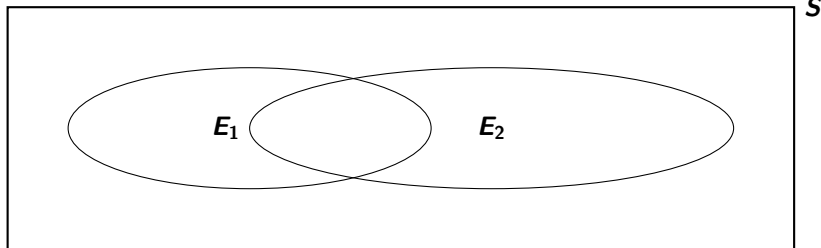
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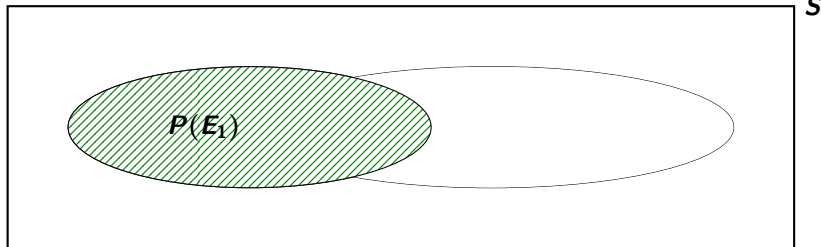
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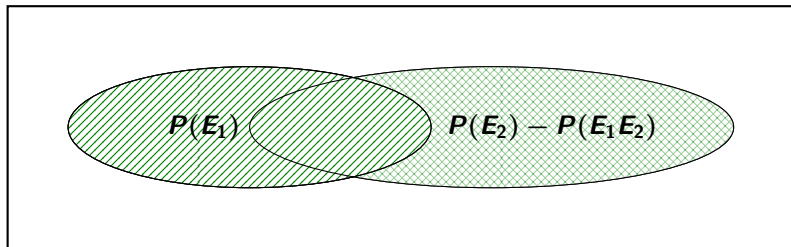
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which yields Axiom 3.

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No. of Vehicles Waiting	No. of Observations	Relative Frequency
0	4	4/60
1	16	16/60
2	20	20/60
3	14	14/60
4	3	3/60
5	2	2/60
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Example 2: Left-turn lane design

60 observations of the number of vehicles waiting for left turns at an intersection are shown in the table.

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0	4	4/60
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Given the following definitions:

$E_1 = \{ > 2 \text{ vehicles waiting for left turns} \}$

$E_2 = \{ \leq 4 \text{ vehicles waiting for left turns} \}$

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Find the following probabilities:

- (a) $P(E_1)$
- (b) $P(E_2)$
- (c) $P(E_1 E_2)$
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Example 2: Left-turn lane design (cont.)

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$$(d) P(E_1 \cup E_2)$$

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How would you characterize E_1 and E_2 ?

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Collectively exhaustive events

Fundamental principle of counting

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Given $1, 2, \dots, k$ operations are to be performed, and n_1, n_2, \dots, n_k ways of performing each respective operation.

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Given $1, 2, \dots, k$ operations are to be performed, and n_1, n_2, \dots, n_k ways of performing each respective operation.

The total number of possibilities is given by

$$n_1 \times n_2 \times \cdots \times n_k$$

Example 3: Ice cream store

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- Three types of cone (cups are not available)
- Ten flavors
- Four toppings (which cannot be combined)

How many distinct ice cream orders are possible?

$$3 \times 10 \times (4 + 1) = \boxed{150}$$

(Note: you can order an ice cream without any topping, hence 5 topping possibilities)

Example 4: License plate numbers

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If in 2021, the state releases a new series of plate numbers that must now end with a letter in addition to 5 preceding digits, how many distinct plates can be manufactured?

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If in 2021, the state releases a new series of plate numbers that must now end with a letter in addition to 5 preceding digits, how many distinct plates can be manufactured? Answer: $26 \times 10^5 = 2,600,000$

Permutations

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Examples of factorials

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Examples of factorials

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- $4! = 4 \times 3 \times 2 \times 1 =$

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- $3! = 3 \times 2 \times 1 = 6$
- $4! = 4 \times 3 \times 2 \times 1 = 24$

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A permutation is an arrangement (ordering) of objects in a collection.

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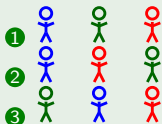
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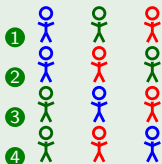
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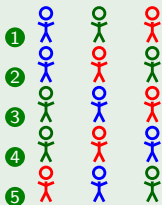
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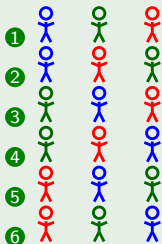
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Permutations with repetition

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Without repetition, the number of permutations of n objects is $n!$.

However, if there are n_1 identical (or repeated) items of type 1, n_2 identical items of type 2, \dots , and n_k identical items of type k , then the number of permutations is given by

$$\frac{n!}{n_1!n_2! \times \dots \times n_k!} \quad (15)$$

How many ways can the letters in the word "MASS" be arranged?

The letter "S" is repeated. Thus, the number of permutations are:

$$\frac{4!}{2!} = 4 \times 3 = 12$$

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With combinations, the order is not important (in contrast to permutations).

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You can form 56 different teams.

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