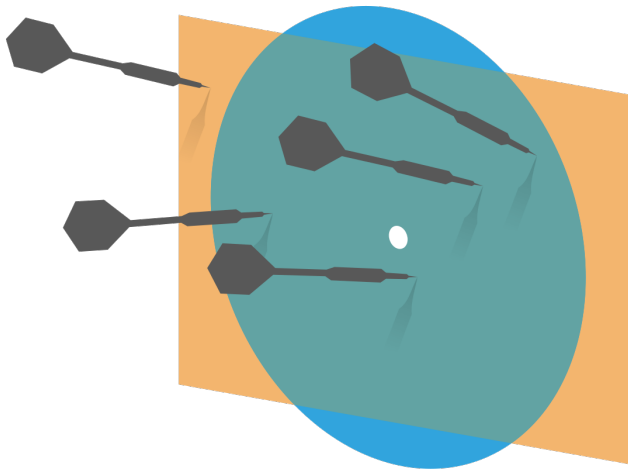


# Probability and Statistics for Computer Science



Credit: wikipedia

“A major use of probability in statistical inference is the updating of probabilities when certain events are observed” – Prof. M.H. DeGroot

# Last time

## Probability

More probability calculation

- Conditional probability

\* Bayes rule ✓

\* Independence

# Objectives

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

## ✱ Conditional Probability

\* Product rule of joint prob.

\* Bayes rule

\* Independence

# Counting: how many ways?

to put 7 hats (hats are  
indistinguishable) on 7 of 10  
people randomly?



$$\binom{10}{7}$$



# Warm up: which is larger?

$$P(A \cap B) \approx P(A|B)$$

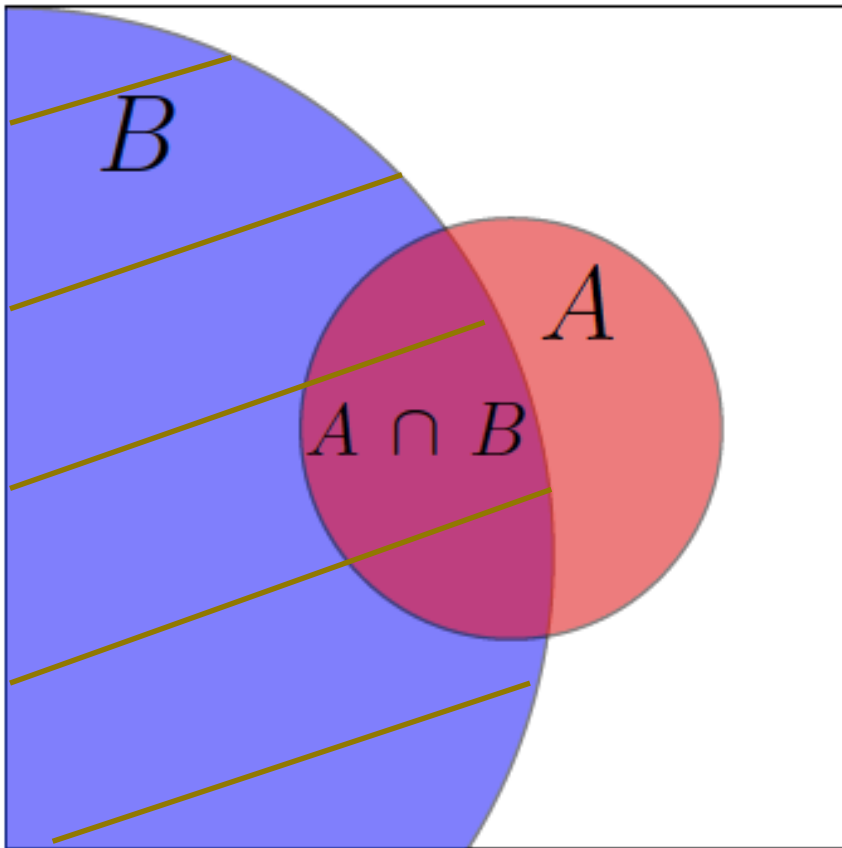
A)  $P(A \cap B)$

B)  $P(A|B)$

C) unsure

# Conditional Probability

✱ The probability of **A** given **B**



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

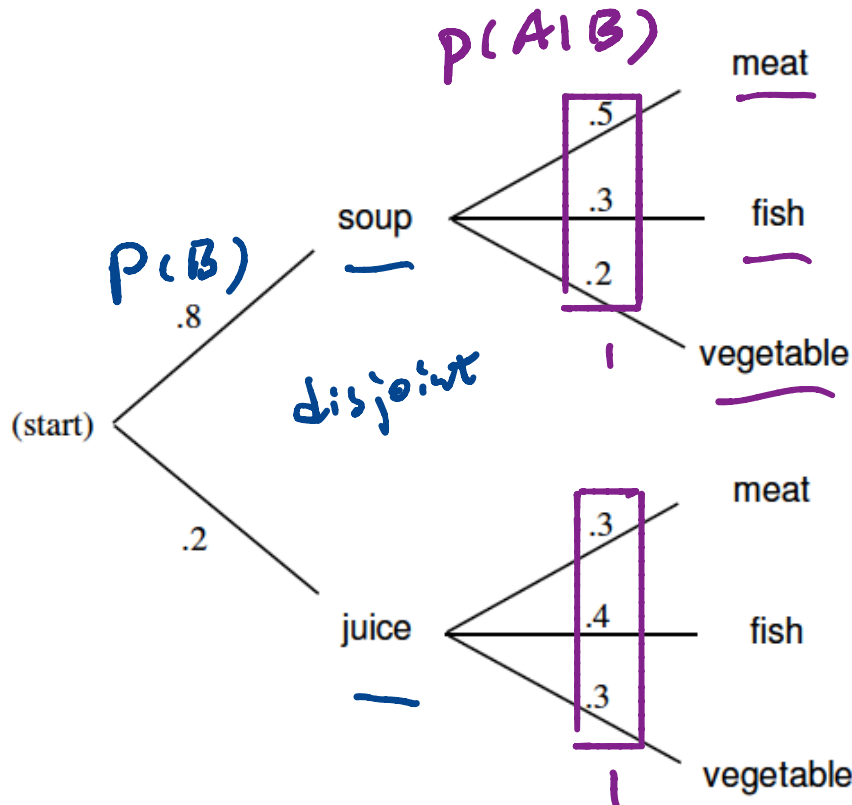
$$P(B) \neq 0$$

The line-crossed area is the new sample space for conditional  $P(A|B)$

# Joint Probability Calculation

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

$$P(A \cap B) \\ 0.8 \times 0.5$$



$$\begin{aligned} P(\text{soup} \cap \text{meat}) &= \\ P(\text{meat}|\text{soup})P(\text{soup}) &= \\ &= 0.5 \times 0.8 = 0.4 \end{aligned}$$

# Bayes rule

- ✱ Given the definition of conditional probability and the symmetry of joint probability, we have:

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

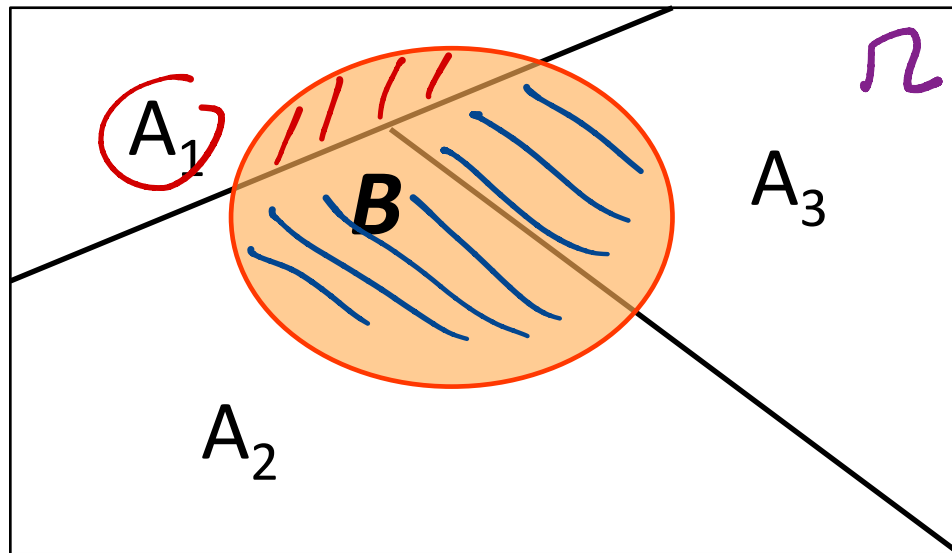
And it leads to the famous Bayes rule:

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

$$\frac{P(A \cap B)}{P(B)}$$

# Total probability

$$\begin{aligned} P(B) &= P(B \cap A) + P(B \cap A^c) \\ &= P(B|A)P(A) + P(B|A^c)P(A^c) \end{aligned}$$

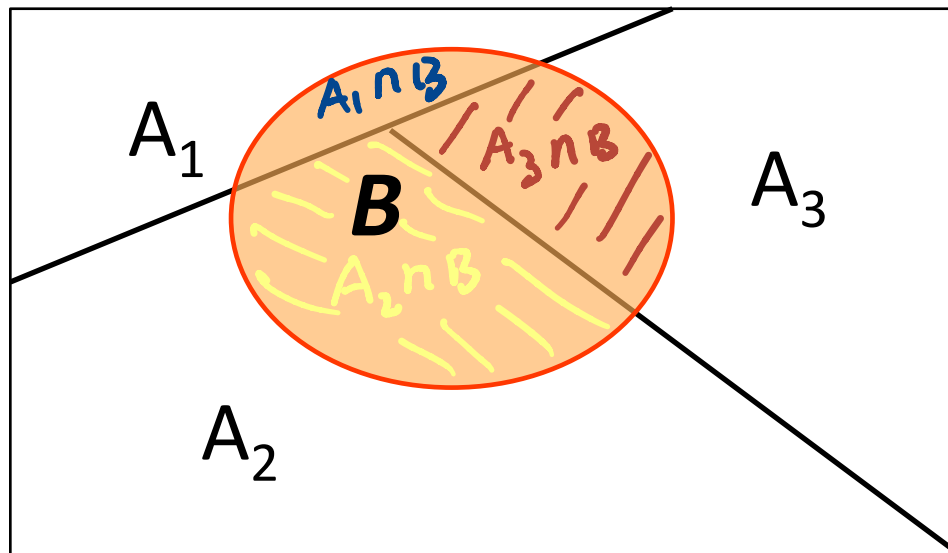


i.e.

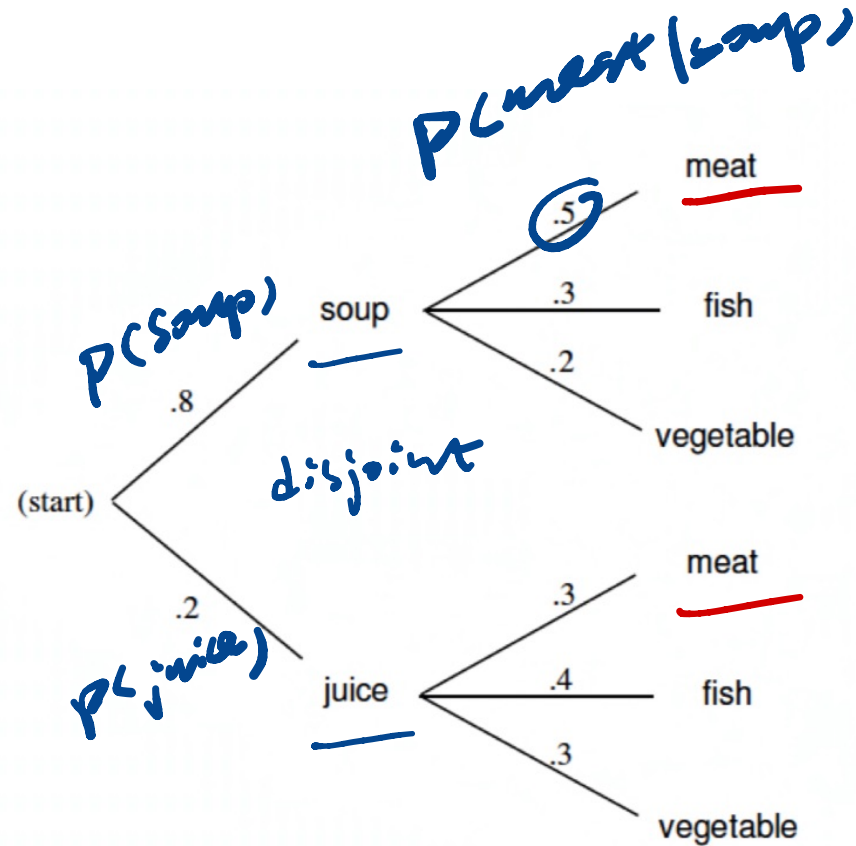
$$\begin{aligned} A &= A_1 \\ A^c &= A_2 \cup A_3 \end{aligned}$$

# Total probability general form

$$\begin{aligned} P(B) &= \sum_j P(B \cap A_j) \\ &= \sum_j P(B | A_j) P(A_j) \end{aligned}$$



# Total probability:



$$P(\text{meat}) = ? \quad P(\text{soup} \cap \text{meat}) + P(\text{juice} \cap \text{meat})$$
$$P(\text{meat} | \text{soup}) P(\text{soup})$$

$$+ P(\text{meat} | \text{juice}) P(\text{juice})$$

$$P(\text{soup} | \text{meat}) = \frac{P(\text{meat} \cap \text{soup})}{P(\text{meat})}$$

# Bayes rule using total prob.

$$P(A_j | B) = \frac{P(B | A_j) P(A_j)}{P(B)}$$

$$\frac{P(A_j \cap B)}{P(B)} = \frac{P(B | A_j) P(A_j)}{\sum_j P(B | A_j) P(A_j)}$$

$$A_j \cap A_i = \emptyset \rightarrow \text{disjoint} \\ \text{if } i \neq j$$



# Bayes rule: rare disease test

There is a blood test for a rare disease. The frequency of the disease is  $1/100,000$ . If one has it, the test confirms it with probability 0.95. If one doesn't have, the test gives false positive with probability 0.001. What is  $P(D|T)$ , the probability of having disease given a positive test result?

$$P(D|T) = \frac{P(T|D)P(D)}{P(T)} \leftarrow \text{Using total prob.}$$
$$= \frac{P(T|D)P(D)}{P(T|D)P(D) + P(T|D^c)P(D^c)}$$

# Bayes rule: rare disease test

There is a blood test for a rare disease. The frequency of the disease is **1/100,000**. If one has it, the test confirms it with probability **0.95**. If one doesn't have, the test gives false positive with probability **0.001**. What is  $P(D|T)$ , the probability of having disease given a positive test result?

$$P(D|T) = \frac{P(T|D)P(D)}{P(T|D)P(D) + P(T|D^c)P(D^c)}$$

$$= \frac{0.95 \times \frac{1}{100,000}}{0.95 \times \frac{1}{100,000} + 0.001 \times \left(1 - \frac{1}{100,000}\right)} \approx < 1\%$$

What about Covid test?

Suppose freq. of Covid = 1.21%

test accuracy = 95%

false positive = 0.001

$P(D|T) = ?$

$$= \frac{P(T|D)P(D)}{P(T|D)P(D) + P(T|D^c)P(D^c)}$$

= 92%

$P(T|D) = 0.95$

$P(D) = 1.21\%$

$P(T|D^c) = 0.001$

$P(D^c) = 1 - 1.21\%$

# Independence

✱ One definition:

$$\begin{aligned} P(A|B) &= P(A) \text{ or} \\ P(B|A) &= P(B) \end{aligned}$$

Whether A happened doesn't change the probability of B and vice versa

# Independence: example

- ✱ Suppose that we have a fair coin and it is tossed twice. let A be the event “the first toss is a head” and B the event “the two outcomes are the same.”

A H\*

B HH TT

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\{HH\}}{\{HH, TT\}} = \frac{1}{2} = P(A) = \frac{1}{2}$$

- ✱ These two events are independent ? ✓

# Independence

## ✱ Alternative definition

LHS by definition  $P(A|B) = P(A)$

$$\Rightarrow \frac{P(A \cap B)}{P(B)} = P(A)$$

$$\Rightarrow P(A \cap B) = P(A)P(B) \quad \checkmark$$

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

# Testing Independence:

- ✱ Suppose you draw one card from a standard deck of cards.  $E_1$  is the event that the card is a King, Queen or Jack.  $E_2$  is the event the card is a Heart. Are  $E_1$  and  $E_2$  independent?

$$P(E_1 \cap E_2) = \frac{3}{52}$$

$$P(E_1) = \frac{3 \times 4}{52} = \frac{3}{13}$$

$$P(E_2) = \frac{1}{4}$$

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

$$\frac{3}{13} \times \frac{1}{4} = \frac{3}{52}$$

# Independence vs Disjoint

✱ Q. Two disjoint events that have probability  $> 0$  are certainly dependent to each other.  $\phi$

A. True

B. False

$$P(A \cap B) = 0$$

$$P(A) > 0 \quad P(B) > 0$$

$$\Rightarrow \underline{P(A \cap B) \neq P(A)P(B)}$$



# Independence of empty event

✱ Q. Any event is independent of empty event B.

**A.** True

B. False

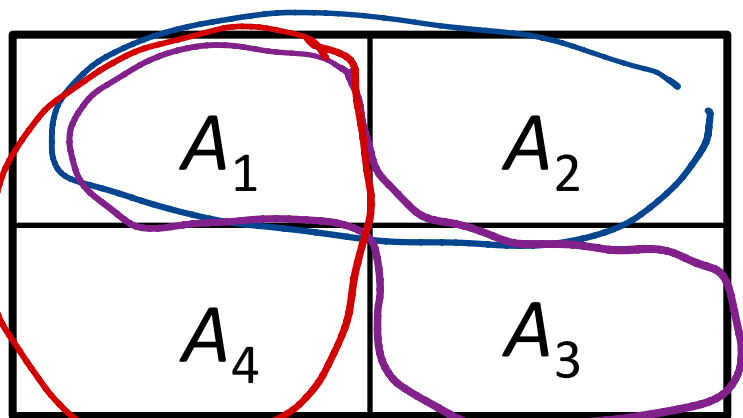
$$B = \phi$$

$$P(\phi) = 0$$

$$P(\phi \cap A) = P(\phi)P(A) = 0$$

$$P(\phi) =$$

# Pairwise independence is not mutual independence in larger context



$$A_1 \cup A_2 \cup A_3 \cup A_4 = \underline{\underline{\Omega}}$$

$$P(A_1) = P(A_2) = P(A_3) = P(A_4) = 1/4$$

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

$$P(B \cap C) = P(B)P(C) \checkmark$$

...

$$A = A_1 \cup A_2; P(A) = 1/2$$

$$B = A_1 \cup A_3; P(B) = 1/2$$

$$C = A_1 \cup A_4; P(C) = 1/2$$

$$P(A \cap B \cap C) = P(A_1) = 1/4$$

\* $P(ABC)$  is the shorthand for  $P(A \cap B \cap C)$

$$P(A)P(B)P(C) = \left(\frac{1}{2}\right)^3$$

if  $(MI) \Rightarrow (PI)$

# Mutual independence

- ✱ Mutual independence of a collection of events  $A_1, A_2, A_3 \dots A_n$  is :

$$P(A_i | A_j \cap A_k \cap \dots \cap A_p) = P(A_i)$$
$$\underline{P(A|B \cap C)} = P(A) \Rightarrow \frac{P(A \cap B \cap C)}{P(B \cap C)} = P(A)$$

↓

- ✱ It's very strong independence!

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

# Probability using the property of Independence: Airline overbooking (1)

- ✱ An airline has a flight with 6 seats. They always sell 7 tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that the flight is overbooked?

$$\begin{aligned} & P^7 \\ \Rightarrow & P(A_1 \cdots A_7) \\ & = P(A_1) P(A_2) \cdots P(A_7) \\ & = p \cdot p \cdots p \end{aligned}$$

$\binom{?}{?} = !$

# Probability using the property of Independence: Airline overbooking (1)

- ✱ An airline has a flight with 6 seats. They always sell 7 tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that the flight is overbooked ?

$P(7 \text{ passengers showed up})$

# Probability using the property of Independence: Airline overbooking (2)

- ✱ An airline has a flight with 6 seats. They always sell 8 tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that exactly 6 people showed up?

$$\text{Event} = \bigcup A_j$$

↑ =

$$(1-p)(1-p)$$

$$p \cdot p \cdot p \cdot p \cdot p \cdot p$$

$$P(6 \text{ people showed up}) =$$

$$= \binom{8}{6} \cdot p^6 (1-p)^2$$

## Probability using the property of Independence: Airline overbooking (3)

- ✱ An airline has a flight with 6 seats. They always sell 8 tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that the flight is overbooked ?

$P(\text{overbooked}) =$

# Probability using the property of Independence: Airline overbooking (4)

- ✱ An airline has a flight with  $s$  seats. They always sell  $t$  ( $t > s$ ) tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that exactly  $u$  people showed up?

P( exactly  $u$  people showed up)

$$\binom{t}{u} \cdot p^u (1-p)^{t-u}$$



# Probability using the property of Independence: Airline overbooking (5)

- ✱ An airline has a flight with  $s$  seats. They always sell  $t$  ( $t > s$ ) tickets for this flight. If ticket holders show up independently with probability  $p$ , what is the probability that the flight is overbooked?

$t > s$

$s$

$\frac{s+1}{t}$

$\dots$

$\frac{s+2}{t}$

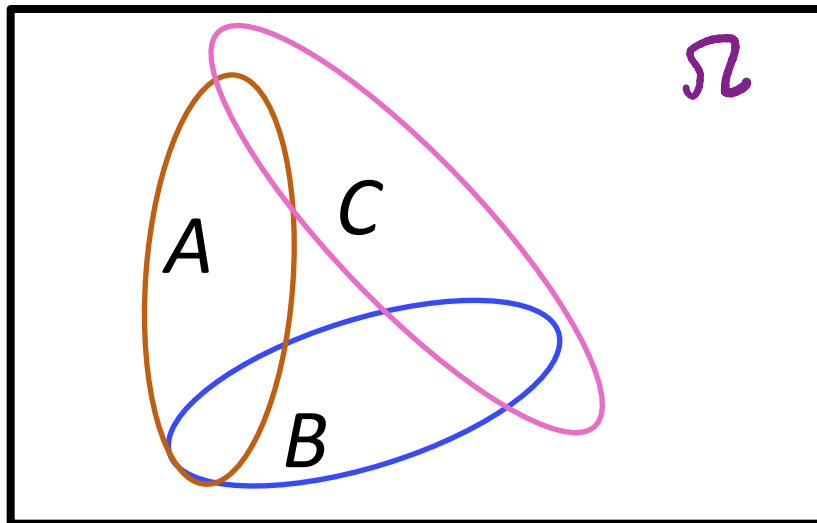
$\dots$

$\frac{t}{t}$

$P(\text{overbooked}) = \sum_{u=s+1}^t \binom{t}{u} p^u (1-p)^{t-u}$

# Condition may affect Independence

- ✱ Assume event **A** and **B** are pairwise independent



Given **C**, **A** and **B** are not independent any more because they become disjoint

# Conditional Independence

✱ Event **A** and **B** are conditional independent given event **C** if the following is true.

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

See an example in Degroot et al. Example 2.2.10

# Assignments

✱ HW3



✱ Finish Chapter 3 of the textbook

✱ Next time: Random variable

# Additional References

- ✱ Charles M. Grinstead and J. Laurie Snell  
"Introduction to Probability"
- ✱ Morris H. Degroot and Mark J. Schervish  
"Probability and Statistics"

# Another counting problem

- ✱ There are several ( $>10$ ) freshmen, sophomores, juniors and seniors in a dormitory. In how many ways can a team of 10 students be chosen to represent the dorm? There are no distinction to make between each individual student other than their year in school.

See you next time

*See  
You!*



