## The Inclusion-Exclusion Principle

## 1. The probability that at least one of two events happens

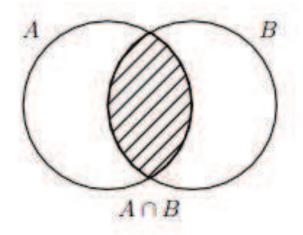
Consider a discrete sample space  $\Omega$ . We define an event A to be any subset of  $\Omega$ , which in set notation is written as  $A \subset \Omega$ . Then, Boas asserts in eq. (3.6) on p. 732 that<sup>1</sup>

$$P(A \cup B) = P(A) + P(B) - P(A \cap B), \qquad (1)$$

for any two events  $A, B \subset \Omega$ . This is equivalent to the set theory result,

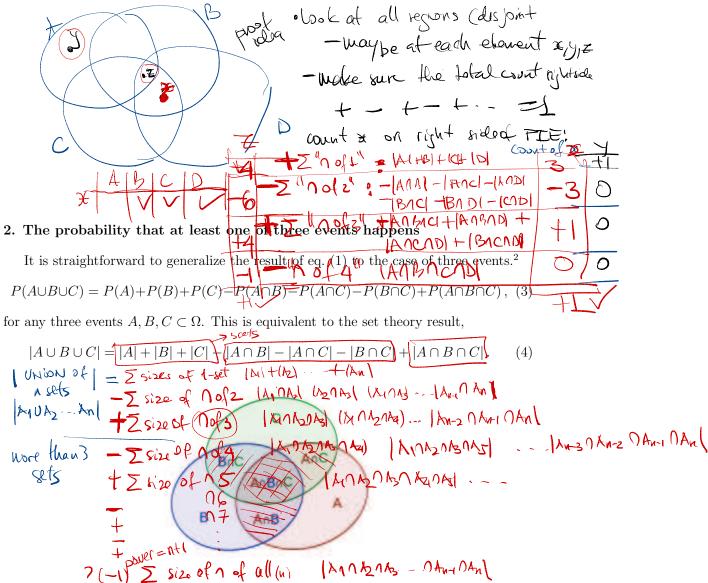
$$|A \cup B| = |A| + |B| - |A \cap B|, \qquad (2)$$

where the notation |A| means the number of elements contained in the set A, etc. In writing eq. (2), we have assumed that A and B are two finite discrete sets, so the number of elements in A and B are finite.



The proof of eq. (2) is immediate after considering the Venn diagram shown above. In particular, adding the number of elements of A and B overcounts the number of elements in  $A \cup B$ , since the events in  $A \cap B$  have been double counted. Thus, we correct this double counting by subtracting the number of elements in  $A \cap B$ , which yields eq. (2). The corresponding result in probability theory is given by eq. (1).

<sup>&</sup>lt;sup>1</sup>Boas uses a nonstandard notation by writing A + B for  $A \cup B$ . The latter is standard in set theory and we shall use it in these notes.  $A \cup B$  means the union of the sets A and B and is equivalent to the "inclusive or," i.e. "either A or B or both." Likewise, Boas uses a nonstandard notation by writing ABfor  $A \cap B$ . Again, the latter is standard in set theory and we shall use it in these notes.  $A \cap B$  means the intersection of the sets A and B, or equivalently "both A and B."



Once again, the proof of eq. (4) is immediate after considering the Venn diagram shown above.<sup>3</sup> In particular, adding the number of elements of A, B and C counts elements in  $A \cap B \cap C$  three times, and counts elements of  $A \cap B$ ,  $A \cap C$  and  $B \cap C$  not contained in  $A \cap B \cap C$  twice. Thus,  $|A \cup B \cup C| = |A| + |B| + |C| - |A \cap B| - |A \cap C| - |B \cap C|$ will include all events in A, B and C once except for the events in  $A \cap B \cap C$ , which were all subtracted off. Thus, to include all events in  $A \cup B \cup C$  exactly once, we must add back the number of events in  $A \cap B \cap C$ . Thus, eq. (4) is established. The corresponding result in probability theory is given by eq. (3).

## 3. The Inclusion-Exclusion principle

The inclusion-exclusion principle is the generalization of eqs. (1) and (2) to n sets. Let  $A_1, A_2, \ldots, A_n$  be a sequence of n events. Then,

$$\underbrace{P(A_1 \cup A_2 \cup \dots \cup A_n)}_{i=1} = \sum_{i=1}^n P(A_i) - \sum_{i < j} P(A_i \cap A_j) + \sum_{i < j < k} P(A_i \cap A_j \cap A_k) \\
- \sum_{i < j < k < \ell} P(A_i \cap A_j \cap A_k \cap A_\ell) + \dots + (-1)^{n+1} P(A_1 \cap A_2 \cap \dots \cap A_n), (5)$$

where  $A_1, A_2, \ldots, A_n \subset \Omega$ . This is equivalent to the set theory result,

$$|A_1 \cup A_2 \cup \dots \cup A_n| = \sum_{i=1}^n |A_i| - \sum_{i < j} |A_i \cap A_j| + \sum_{i < j < k} |A_i \cap A_j \cap A_k|$$
$$- \sum_{i < j < k < \ell} |A_i \cap A_j \cap A_k \cap A_\ell| + \dots + (-1)^{n+1} |A_1 \cap A_2 \cap \dots \cap A_n|.$$
(6)

The proof of eq. (6) is an exercise in counting. Suppose a point is contained in exactly m of the sets,  $A_1, A_2, \ldots, A_n$ , where m is a number between 1 and n. Then, the point is counted m times in  $\sum_{i=1}^{n} |A_i|$ , it is counted C(m, 2) times in  $\sum_{i<j} |A_i \cap A_j|$ , it is counted C(m, 3) times in  $\sum_{i<j< k} |A_i \cap A_j \cap A_k|$ , etc., where C(m, k) is the number of combinations of m objects taken k at a time. After reaching  $\sum_{i_1 < i_2 < \cdots i_m} |A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_m}|$ , where the point is counted once [since C(m, m) = 1], one finds that the point is not counted at all in any of the terms that involve the intersection of more than m sets. The net result is that a point that is contained in exactly m of the sets will be counted S times in  $|A_1 \cup A_2 \cup \cdots \cup A_n|$  given by eq. (6), where

$$S \equiv C(m,1) - C(m,2) + C(m,3) - C(m,4) + \dots + (-1)^{m+1}C(m,m),$$
(7)

after noting that C(m, 1) = m.

To compute S, we recall the binomial theorem,

$$(x+y)^m = \sum_{k=0}^m C(m,k) x^k y^{m-k},$$
(8)

where

$$C(m,k) \equiv \binom{m}{k} \equiv \frac{m!}{k!(m-k)!}$$

is the number of combinations of m objects taken k at a time. Setting x = 1 and y = -1 in eq. (8) yields,

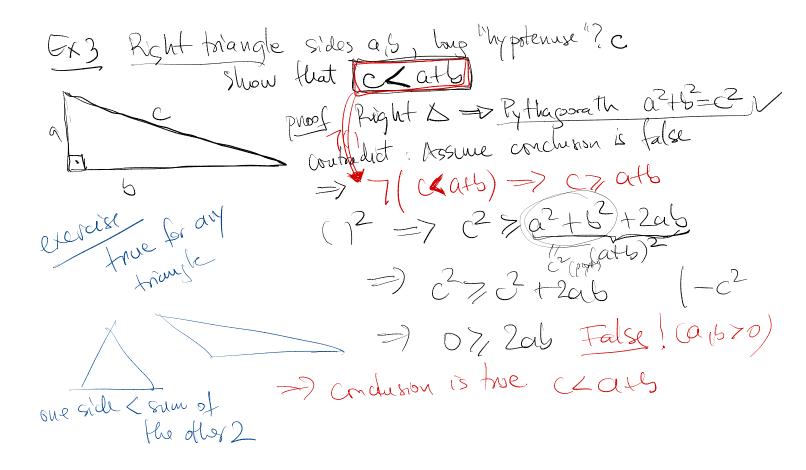
$$\sum_{k=0}^{m} (-1)^k C(m,k) = 0.$$

Using C(m, 0) = 1, it follows that

$$1 - C(m, 1) + C(m, 2) - C(m, 3) + \ldots + (-1)^m C(m, m) = 0$$

which implies that S = 1 [cf. eq. (7)]. Thus, we have shown that there is no multiple counting of points in eq. (6). That is, every point contained in the union of  $A_1, A_2, \ldots, A_n$ is counted exactly one time. Thus, eq. (6) is established. The corresponding result in probability theory is given by eq. (5). We have therefore verified the inclusion-exclusion principle.

There are numerous applications of the inclusion-exclusion principle, both in set theory and in probability theory. In particular, it provides a powerful tool for certain types of counting problems. An example is provided in the next section of these notes.



part A, Satisfiability Intro [easy]. A boolean formula is satisfiable if there exists some variable assignment that makes the formula evaluate to true. Namely, a boolean formula is satisfiable if there is some row of the truth table that comes out true. Determining whether an arbitrary boolean formula is satisfiable is called the *Satisfiability Problem*. There is no known efficient solution to this problem, in fact, an efficient solution would earn you a million dollar prize. While this is hard problem in computer science, not all instances of the problem are hard, in fact, determining satisfiability for some types of boolean formulae is easy.

- i. First, let's consider why this would be hard. If you knew nothing about a given boolean formula other than that it had n variables, how large is the truth table you would need to construct? Please indicate the number of columns and rows as a function of n
- ii. Now consider the following 100 variable formula.

$$x_1 \wedge (\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_3) \wedge (\neg x_3 \vee x_4) \wedge \ldots \wedge (\neg x_{99} \vee x_{100})$$

Without constructing a truth table, how many satisfying assignments does this formula have, explain your answer.

iii Now consider an arbitrary 3-DNF formula with 100 variables and 200 clauses. 3-DNF means that the formula is in disjunctive normal form and each clause has three literals. (A literal is the instantiation of the variable in the formula, so for x,  $\neg x$  or x.) An example might be something like:

$$(\neg x_1 \land x_3 \land x_{10}) \lor (\neg x_3 \land x_{15} \land \neg x_{84}) \lor (x_{17} \land \neg x_{37} \land x_{48}) \lor \ldots \lor (\neg x_{87} \land \neg x_{95} \land x_{100})$$

What is the largest size truth table needed to solve this problem. What is the maximum number of such truth tables needed to determine satisfiability.

part B: 2CNF-SAT [hard]. The 2CNF-SAT instance is a boolean CNF formula with 2 variables in each clause, "OR" inside clauses, "AND" between clauses. There are m boolean variables  $x_1, x_2, ..., x_m$  and n clauses  $C_1, C_2, ..., C_n$ . Every variable and its negation appears in at least one clause. Such formula is given as input in Now to think about 2 CNF Formulas format redundantly : -every clause can be written - for each variable there is a list of clauses containing it as z'implications (redundant) - for each clause there there are 2 variables For example the formula  $x_1 \vee \neg x_2 \wedge$  $\vee x_3$  $\neg x_2 \lor \neg x_3$  will be  $(x_2 \vee x_3)$  $x_1$ given as: ヨフトろ (X) 722 12=77x3 m = 3, n = 4X2 => 7×2  $x_1 : C_1$ -draw these implication (not noc, but see ful)  $\neg x_1 : C_3$ • that and error start like = T  $x_2: C_2$  $\neg x_2: C_1, C_4$ for example try X2=T => X1=E => X3=E.  $x_3: C_2, C_3$  $\neg x_3 : C_4$  $C_1: x_1, \neg x_2$  $C_2: x_2, x_2$  $C_3: \neg x_1, x_3$  $C_4: \neg x_2, \neg x_3$ 

Your task is to design a strategy that determines, for a given formula, the boolean assignments for the variables such that all clauses are satisfied, thus the formula is true (if more such assignments are possible, you only need to output one). If no such assignment is possible, output "FALSE".

As established inpart A, there are  $2^m$  possible assignments for the variable set. So if one were to build the truth table and "brute force" search all rows/assignments until one works, it would take exponential time — not good! Instead: do trial and error, but in a smart way that only tries at most  $2 * m^2$  boolean assignments.

Your strategy can be pseudocode, or you can informally describe a procedure with bullets and English statements. You can write in your procedure statements like x = x

* $x = x_1$	ZENT		
* foreach $C$ containing variable $x$ {	$(x_2 \vee 7 x_3) \Lambda($	$(\times_3, \vee, \times_4) \land ()$	$(x_2 \sqrt{1} x_1) \wedge (x_3 \sqrt{1} x_2)$
	13=> 1/2	`	5
} * $C$ = next clause, or $C$ = next claus	$[\chi_2 \rightarrow 1_{3}]$	$(\tilde{A})$	
C = field clause, of $C = field$ claus	e containing x		(13)
* loop C through all clauses that cor	tain $x$ or $\neg x$	6	
* for each $x \in C$ {		(74)	$1 (7_{y_2})$
 }		$\bigcirc$	(42)
* $y =$ the other variable in clause C, other than x or $\neg x$			