

JPL-Caltech Virtual Summer School

Big Data Analytics

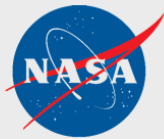
September 2 – 12, 2014

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Basic Probability

Part 1



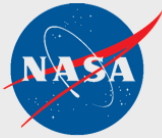
National Aeronautics and
Space Administration

Jet Propulsion Laboratory
California Institute of Technology
Pasadena, California

Outline

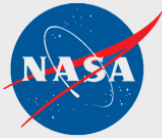
Introduce basic concepts of probability and some mathematical machinery:

- ▶ What is probability?
- ▶ Sample spaces and events.
- ▶ Axioms of Probability and some corollaries.
- ▶ Joint and conditional probabilities.



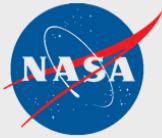
What is probability?

- ▶ We all have an idea of what probability is:
 - ▶ the probability it will rain today,
 - ▶ the probability the freeway is jammed,
 - ▶ the probability I will roll double sixes,
 - ▶ the probability global warming is real.
- ▶ Probability can be defined as long-run relative frequency.
 - ▶ I flip a coin. What is the probability I get a head?
 - ▶ Answer: $1/2$ since physics tells me that if I flipped the coin 1000 times I should get about 500 heads; 10,000 times I should get 5000 heads, etc.
- ▶ Mathematically, probability is a set function: it assigns a number (between zero and one, in this case) to a set.



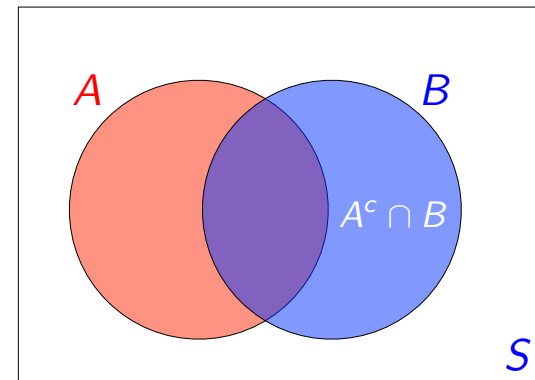
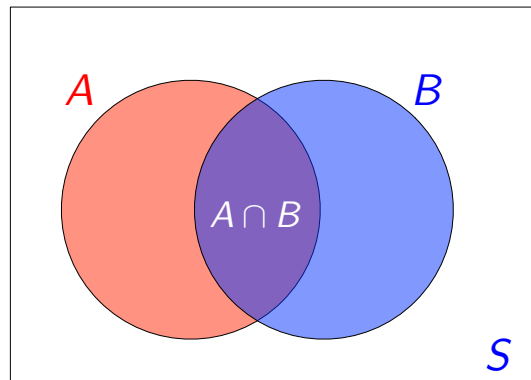
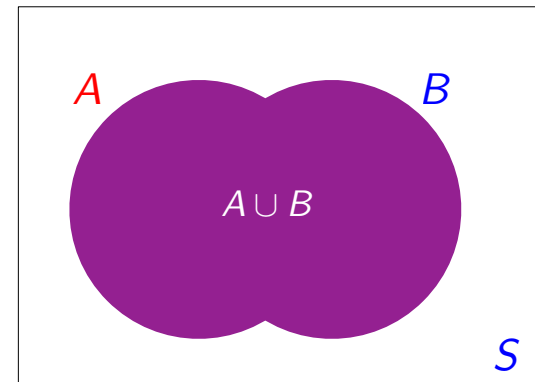
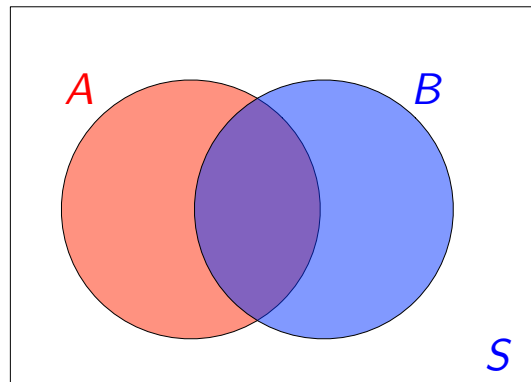
Sample spaces and events

- ▶ Define a trial to be a measurement or observation of some phenomenon.
- ▶ The set of all possible outcomes of a trial is called the sample space, S .
 - ▶ sex of a newborn baby: $S = \{\text{boy}, \text{girl}\}$,
 - ▶ minutes to wait to get on the freeway this morning: $S = \{x : 0 \leq x < \infty\}$.
- ▶ An event is a subset of the sample space.
 - ▶ $E = \{\text{boy}\}$,
 - ▶ five minutes or less: $E = \{x : 0 \leq x \leq 5\}$.

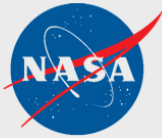


Sample spaces and events

Venn diagrams:



\cap = intersection; \cup = union; c = complement.



Sample spaces and events

Handy “laws” of set operations:

Commutative:

$$A \cup B = B \cup A$$

$$A \cap B = B \cap A$$

Associative:

$$(A \cup B) \cup C = A \cup (B \cup C)$$

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

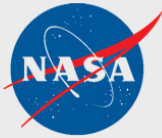
Distributive:

$$(A \cup B) \cap C = (A \cap C) \cup (B \cap C)$$

$$(A \cap B) \cup C = (A \cup C) \cap (B \cup C)$$

DeMorgan’s laws:

$$\left(\bigcup_{i=1}^n A_i \right)^C = \bigcap_{i=1}^n A_i^C, \quad \left(\bigcap_{i=1}^n A_i \right)^C = \bigcup_{i=1}^n A_i^C.$$



Axioms of Probability and some corollaries

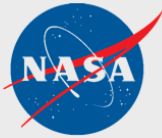
Three axioms:

Axiom 1: $0 \leq P(A) \leq 1.$

Axiom 2: $P(S) = 1.$

Axiom 3: $P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i),$ for a sequence of mutually exclusive events, A_1, A_2, \dots ($A_i \cap A_j = \emptyset$ for $i \neq j$).

All other rules of probability can be derived from these axioms.



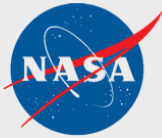
Axioms of Probability and some corollaries

Useful:

- ▶ $P(A^c) = 1 - P(A)$.
- ▶ If $A \subset B$ (A is a subset of B), then $P(A) \leq P(B)$.
- ▶ $P(A \cup B) = P(A) + P(B) - P(A \cap B)$.
- ▶ More generally,

$$\begin{aligned} P(A_1 \cup A_2 \cup \dots \cup A_n) = & \sum_{i=1}^n P(A_i) - \sum_{i_1 < i_2} P(A_{i_1} \cap A_{i_2}) + \dots \\ & + (-1)^{r+1} \sum_{i_1 < i_2 < \dots < i_r} P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_r}) \\ & + \dots + (-1)^{n+1} P(A_1 \cap A_2 \cap \dots \cap A_n), \end{aligned}$$

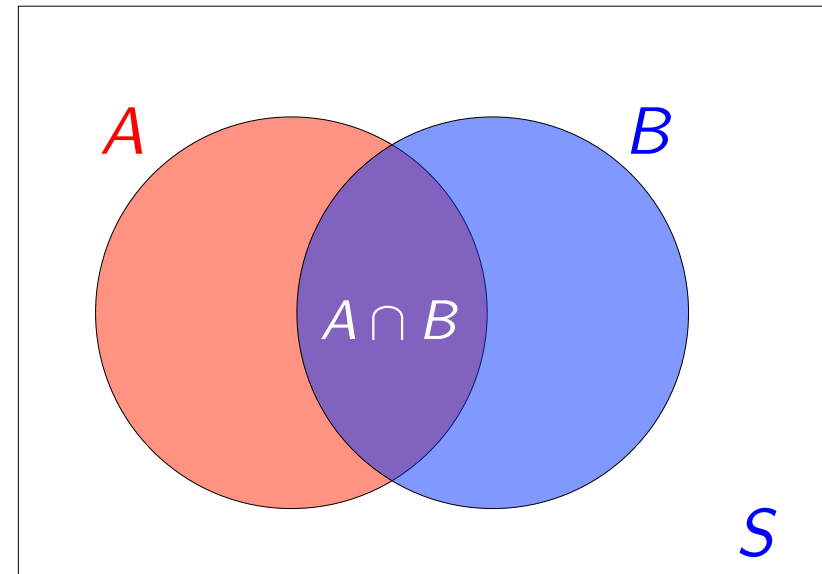
where the summation on the second line is over all possible subsets of size r from the set of size n .



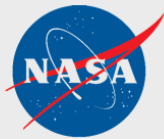
Joint and conditional probabilities

- ▶ The joint probability of events A and B is $P(A \cap B)$.
- ▶ The conditional probability of A given that event B occurs is,

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



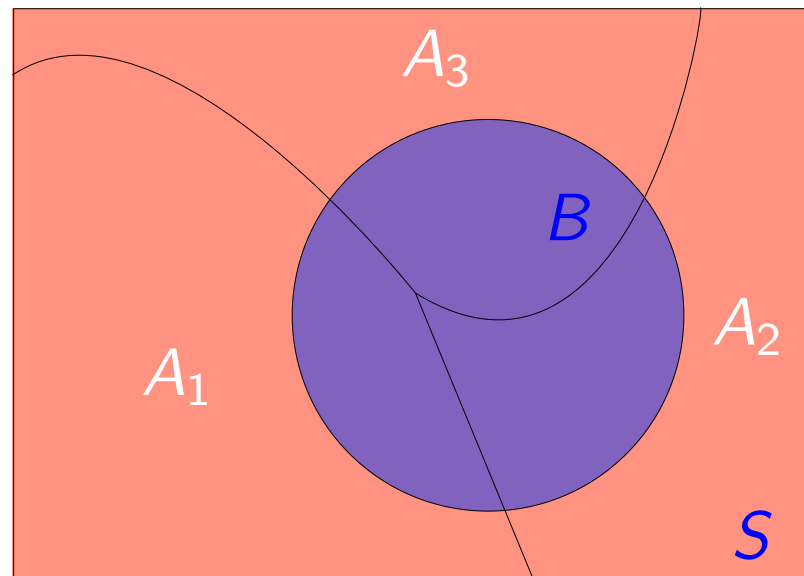
- ▶ The unconditional probability of A is $(P(A \cap S)/P(S))$.
- ▶ Definition of independence: $P(A \cap B) = P(A)P(B)$.
- ▶ Another equivalent definition of independence: $P(A|B) = P(A)$.

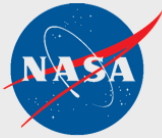


Joint and conditional probabilities

Law of Total Probability:

$$P(B) = \sum_{i=1}^n P(B \cap A_i) = \sum_{i=1}^n P(B|A_i)P(A_i).$$





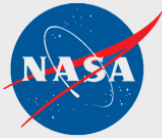
Joint and conditional probabilities

Bayes' Rule:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} = \frac{P(B|A)P(A)}{\sum_{i=1}^n P(B|A_i)P(A_i)}.$$

Bayes' Rule is especially useful because it allows us to express $P(A|B)$ in terms of $P(B|A)$, if we know something about the latter but not the former.

Example: Let A be the event that true CO₂ concentration is greater than 400 ppm. Let B be the event that OCO-2 observes 398 ppm.

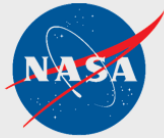


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Bibliography

- ▶ *A First Course in Probability* by Sheldon Ross, Prentice Hall, 2010.
- ▶ *An Introduction to Probability Theory and Its Applications* Volumes 1 and 2, by William Feller, John Wiley and Sons, 1957.



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Next

In the next module, we will discuss how these rules translate for settings in which we model numerical phenomena. In other words: data.