## Rules of Probability Models for Socio-Environmental Data

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#### Road map for today

- Rules of probability
- Factoring joint probabilities
- Directed acyclic graphs (a.k.a. Bayesian networks)

## All of Bayesian inference extends from three rules of probability

- Conditional probability (and independence)
- The law of total probability
- The chain rule of probability

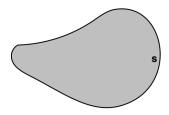
#### Random variables

The world can be divided into things that are observed and things that are unobserved.

- Bayesians treat all unobserved quantities as random variables.
- The values of random variables are governed by chance.
- 3 Probability distributions describe "governed by chance."
- A specific value of a random variable is called an event or an outcome.

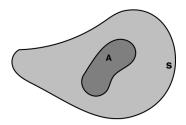
### S=Sample Space

- The set of all possible values of a random variable.
- The sample space, S has a specific area.

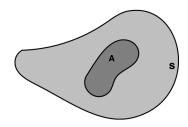


#### Events in S

- Can define and event, A.
- The area of event A is less than or equal to S.



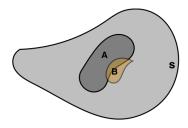
#### What is the probability of event A?



 $\Pr(A) = \frac{\text{Area of } A}{\text{Area of } S}$ 

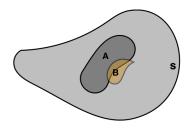
## Conditional Probability

Conditional probability: the probability of an event given that we know another event has occurred.



## Conditional Probability

What is the probability of event B, given that event A has occurred?

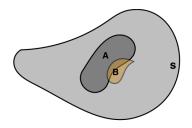


Pr(B|A) = probability of B conditional on knowing A has occurred

$$Pr(B|A) = \frac{\text{Joint Probability}}{\text{Probability of A}} = \frac{Pr(A,B)}{Pr(A)}$$

## Conditional Probability

What is the probability of event A, given that event B has occurred?



#### Independence

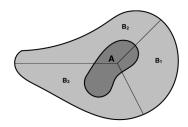
Event A and B are *independent* If the occurrence of event A does not tell us anything about event B.

Events are independent if and only if:

$$Pr(A|B) = Pr(A)$$

$$Pr(B|A) = Pr(B)$$

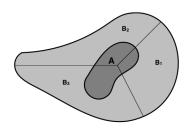
## The Law of Total Probability



Pr(A) is unknown, but can be calculated using the known probabilities of several related events under various conditions.

We can define a set of events  $\{B_n : n = 1, 2, 3, ...\}$ , which taken together define the entire sample space,  $\sum_n B_n = S$ .

## What is the probability of event A?



$$Pr(A) = \sum_{n} Pr(A|B_n) Pr(B_n)$$
 (discrete case)

$$Pr(A) = \int Pr(A|B) Pr(B) dB$$
 (continuous case)

## Intution about the chain rule of probability

Move to course Jamboard

## The Chain Rule of Probability

The chain rule of probability allows us to write joint distributions as a product of conditional distributions.

$$\Pr(z_1, z_2, ..., z_n) = \Pr(z_n | z_{n-1}, ..., z_1) ... \Pr(z_3 | z_2, z_1) \Pr(z_2 | z_1) \Pr(z_1)$$

Notice the pattern here.

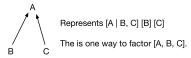
- z's can be scalars or vectors.
- Sequence of conditioning does not matter.
- When we build models, we choose a sequence that makes sense.

#### Factoring joint probabilities

#### Why is factoring useful?

- Factoring joint distributions is how we build Bayesian models.
- The rules of probability allow us to simplify complicated joint. distributions, breaking them down into chunks.
- Chunks can be analyzed one at a time.

# Consider a factored joint distribution represented by a directed acyclic graph (DAG)



- Directed acyclic graphs (aka Bayesian networks) specify how joint distributions are factored into conditional distributions using nodes to represent RV's and arrows to represent dependencies.
- Nodes at the heads of arrows must be on the left hand side of the conditioning symbols;
- Nodes at the tails of arrows are on the right hand side of the conditioning symbols.
- Any node at the tail of an arrow without an arrow leading into it must be expressed unconditionally.
- Nodes at heads of arrows are called "children"; at tails, "parents."

### Factoring joint probabilities using DAGs

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#### Work on lab

Complete Probability Lab #1