# **Lecture 1: Conditional Probability**

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Consider the dreaded disease **Dipsidoodleitis**. Dipsidoodleitis is detected by a blood test which measures dipsidoodlamines in the blood. Values for people without the disease are normally distributed with M=70 and sd=10. Value for people with the disease are M=84, sd=14. Baserates are know as follows: 10% of people have the disease.

You happen to have a dipsidoodleamine value of 88. What is the probability that you are diseas free?

- This is a workshop on doing, meaning generating, solving, programming
- I will not lecture much

# **Probability**

#### Two Domains:

- Rules
- Soul (yes, probability has a soul)

# Rules of Probability, Kolmogorov Axioms

Let X be the sample space (it'a set). Let  $A, B \subseteq X$  be events in X.

- **1.**  $Pr(A) \geq 0$ ,
- **2.** Pr(X) = 1, and
- **3.** if  $A \cap B = \emptyset$ , then  $Pr(A \cup B) = Pr(A) + Pr(B)$ .

#### **Your Turn**

Suppose we flip a coin twice. Here are the probabilities on the outcomes.  $P(\{(H,H)\}) = .5$ ,  $P(\{(H,T)\}) = .2$ ,  $P(\{(T,H)\}) = .15$ . Use the Kolmogov axioms to find:

- 1. The probabilities on all events.
- 2. Probability of at least one tail
- 3. Probability of a trick coin

# **Probability**

- Rules describe relative weight
- Works say on the weight of my body divided into overlapping parts
- But, we need more

### Soul of Probability

### Frequentist (Conventional):

- Probability is proportion in the long run
- Probability is a property of the natural world
  - Probability of coin is a property of the coin, much like its radius, weight, or alloy composition.
- Probability cannot be changed
- Probability is only knowable exactly in the long run
- Probability is not placed on unknown values that are fixed, say a parameter or a model.

### Soul of Probability:

### Bayesian

- Probability is degree of plausibility
- Probability is subjective
- Probability is our mental construct, not part of the world per se
- Probabilities may be placed on parameters, models, individual coin flips, etc.

# **Updating Probability?**

There are three horses in a race. Each has a 1/3rd chance of winning. We are the house and we are going to let people make bets. Each bet costs \$1, and if you pick the winning horse, the bet pays \$2.90. We make 10 cents per bet. Now, right before betting opens, Horse 3 eats a bad apple, gets sick, and withdraws. How should we adjust probability and bets so that we make 10 cents per bet?

# **Bayesian Goal:**

- Treat beliefs as probabilities
- Update beliefs rationally in light of data using rules of conditional probability

### **Your Turn**

• What is a random variable?

### My Turn

Suppose X=(rain,sun,snow) with

- $Pr(\{rain\})=.8$
- $Pr(\{sun\})=.19$
- $Pr(\{snow\})=.01$

### Suppose:

- $Y(\{rain\})=-1$
- $Y(\{sun\})=1$
- $Y(\{snow\})=5$

#### Then

- General:  $Pr(Y = y) = Pr(Y^{-1}(y))$
- Example:  $Pr(Y = 1) = Pr(\{sun\}) = .19$
- Example:  $Pr(Y > 0) = Pr(\{sun, snow\}) = .2$

### My Turn

- Technical A random variable maps from one probability space to another.
- A random variable maps from the events in the world into real numbers. Probabilities on numbers correspond to probabilities on events in the world.

# **Probability Mass function**

$$Pr(Y = y) = \begin{cases} .8, & y = -1, \\ .19, & y = 1, \\ .01, & y = 5, \\ 0, & \text{otherwise.} \end{cases}$$

We can also write it as  $Pr_Y(y)$  or Pr(y), or sometimes, f(y).

#### **Your Turn**

- The above random variables were discrete.
- What is a continuous random variable?
- What is a density function?
- Draw an example for weight of people
- What is Pr(W=60 kg)?
- What is the unit on the *y*-axis?

# Joint Probability

- Foundational, Starting Point
- Example: Two coin flips.
  - $X_1 = 0, 1$ , first flip
  - $X_2 = 0, 1$ , second flip
- Joint
  - $Pr(X_1 = x_1 \cap X_2 = x_2) = Pr(X_1 = x_1, X_2 = x_2) = Pr(x_1, x_2)$
  - Pr(0,0) = .2
  - Pr(0,1) = .3
  - Pr(1,0) = .4
  - Pr(1,1) = .1
- How many "free" probabilities here? 3
- Why, they sum to 1.0

### **Your Turn**

- $Pr(X_1 = 0)$ ,  $Pr(X_1 = 1)$ ,  $Pr(X_2 = 0)$ ,  $Pr(X_2 = 1)$  are called marginal probabilities
- What is Pr(X\_1=0)?
- How do you know this? What is the Law?

# Law of Total Probability

Suppose there is a sequence of sets  $B_1, B_2, \ldots, B_J$  that partitions sample space X ( $X \subseteq \bigcup_j B_j, B_i \cap B_j = \emptyset, i \neq j$ )

$$Pr(A) = Pr(A \cap B_1) + \ldots + Pr(A \cap B_J) = \sum_j Pr(A \cap B_j)$$

• 
$$Pr(X_1 = 0) = Pr(X_1 = 0, X_2 = 0) + Pr(X_1 = 0, X_2 = 1) = .2 + .3 = .5$$

#### Your turn

- Consider marginals  $Pr(X_1 = 0)$ ,  $Pr(X_1 = 1)$ ,  $Pr(X_2 = 0)$ ,  $Pr(X_2 = 1)$ .
- There are 4 in total. How many "free" marginal probabilities are there?
- What are the sums-to-one constraints?

# **Defintion of Conditional Probability**

$$Pr(X_1|X_2) = \frac{Pr(X_1 \cap X_2)}{Pr(X_2)}$$

also

$$Pr(X_1 \cap X_2) = Pr(X_1|X_2)Pr(X_2)$$

### **Your Turn**

- $Pr(X_1 = 1)$
- $Pr(X_1 = 1, X_2 = 0)$
- $Pr(X_1 = 1 | X_2 = 0)$
- How many free parameters in  $Pr(X_1|X_2)$ ?

# Law of Total Probability (conditional format)

$$P(A) = \sum_{i} P(A \cap B_i) = \sum_{i} P(A|B_i)P(B_i)$$

We will use this form a lot soon

# Law of Conditional Probability

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

#### Bernoulli

Frequentist:

$$Y(\theta) \sim \mathsf{Bernoulli}(\theta)$$

Bayesian:

$$Y|\theta \sim \mathsf{Bernoulli}(\theta)$$

Short Cut:

$$Y \sim \mathsf{Bernoulli}(\theta)$$

- The Bernoulli is a discrete distribution, it lives on two points: y = 0 and y = 1.
- The Bernoulli describes dichotomous events, like coin flips.
- The Bernoulli has a single parameter, the probability of success.

### Probabilities for Bernoulli

Frequentist:

$$Pr(Y = y, \theta) = \theta^{y} (1 - \theta)^{(1-y)}, \quad y = 0, 1$$

Bayesian:

$$Pr(Y = y \mid \theta) = \theta^{y} (1 - \theta)^{(1-y)}, \quad y = 0, 1$$

### **Normal Distribution**

Everyone knows the normal.

Freq.

$$Y(\mu, \sigma^2) \sim \mathsf{Normal}(\mu, \sigma^2)$$

Bayes.

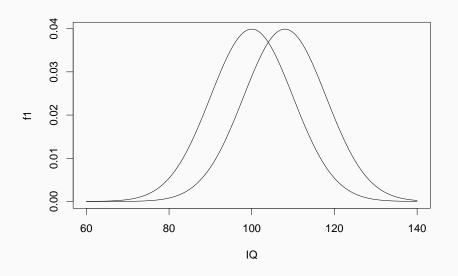
$$Y|\mu,\sigma^2 \sim \mathsf{Normal}(\mu,\sigma^2)$$

Short Cut.

$$Y \sim \mathsf{Normal}(\mu, \sigma^2)$$

$$f(y|\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(y-\mu)^2}{2\sigma^2}\right)$$

### **Normal Distribution**



#### **Normal Distribution**

- 1. The normal is a continuous distribution, it lives on the interval  $(-\infty,\infty)$
- 2. The normal has a location parameter  $\mu$  and a variance parameter  $\sigma^2$
- 3. The normal is used because of its convenience.
- **4.** Effects shift the distribution.

# Working the Problem

Consider the dreaded disease **Dipsidoodleitis**. Dipsidoodleitis is detected by a blood test which measures dipsidoodlamines in the blood. Values for people without the disease are normally distributed with M=70 and sd=10. Value for people with the disease are M=84, sd=14. Baserates are know as follows: 10% of people have the disease.

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# Working The Problem

**Notation Step:** Let Let Y denote be a random variable that denotes the dipsidoodlamine level. Let D be a random variable that denotes whether an individual has the disease or not. (D=0,1 for no disease and disease, respectively).

**Model Step** What kind of model statements can we make? Are they conditional or marginal statements?

- $D \sim \text{Bernoulli}(.1)$
- $Y|D = 0 \sim \text{Normal}(70, 10^2)$
- $Y|D = 1 \sim \text{Normal}(84, 14^2)$