

# Lecture 7: Decision Trees

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*Based on Prof. Rishabh Iyer's slides.*

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## Part I

# Probability, Random Variables, and Entropy

## 1 Probability

The following content is sourced from the following:

- [seeing-theory](#)
- [Khan Academy](#)
- Professor Rishabh Iyer's class notes

### 1.1 Discrete Probability

#### Definition 1.1 Sample space

**Sample space** specifies the set of possible outcomes.

For example,  $\Omega = \{H, T\}$  would be the set of possible outcomes of a coin flip. Simply put, we are saying a coin flip could be either *heads* or *tails*.

#### Definition 1.2 Probability

**Probability** is simply how likely something is to happen

For each element  $\omega$  inside of our sample space  $\Omega$ , there is a number  $p(\omega) \in [0, 1]$  called a probability. This represents how likely the event  $\omega$  is to happen.

The total probability of all possible outcomes (each outcome being an element  $\omega$ ) within the sample space  $\Omega$  adds up to 1. This means that one of the outcomes in the sample space is certain to occur.

$$\sum_{\omega \in \Omega} p(\omega) = 1$$

For example, a biased coin might have  $p(H) = .6$  and  $p(T) = .4$ .

Note:  $[0, 1]$  is a range, not a set. This number is always between 0 and 1, where 0 indicates impossibility and 1 indicates certainty.

### Definition 1.3 Event

An **event** is a subset of the sample space  $\Omega$ , e.g.

- Let  $\Omega = \{1, 2, 3, 4, 5, 6\}$  be the 6 possible outcomes of a dice roll.
- $A = \{1, 5, 6\} \subseteq \Omega$  would be the event that the dice roll comes up as a one, five, or six.

The probability of an event is just the sum of all the outcomes that it contains.

$$p(A) = p(1) + p(5) + p(6)$$

Note:  $A \subseteq \Omega$  means that  $A$  is a subset of  $\Omega$ . Below, is a brief recap of what a subset is.

### Definition 1.4 Subset

A set  $A$  is a **subset** of another set  $B$  if all of elements of the set  $A$  are elements of the set  $B$ . In other words, the set  $A$  is contained inside the set  $B$ .

## 1.2 Conditional Probability

### 1.2.1 Independence

We say two events are independent if knowing one event occurred doesn't change the probability of the other event. For example, the probability that a fair coin shows "heads" after being flipped is  $\frac{1}{2}$ . What if we knew the day was Tuesday? Does this change the probability of getting "heads"? Of course not. The probability of getting "heads", given that it's a Tuesday, is still  $\frac{1}{2}$ . Thus, the result of a coin flip and the day being Tuesday are independent events. [More reading](#).

### Definition 1.5 Independence

Two events  $A$  and  $B$  are independent if

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

Informally,  $p(A \cap B)$  indicates the probability of  $A$  and  $B$  (aka probability of  $A$  intersection  $B$ ). This means that it indicates the likelihood of both events happening.

Let's suppose that we have a fair die:  $p(1) = \dots = p(6) = \frac{1}{6}$  (the probability of rolling any number on the dice is equal). If  $A = \{1, 2, 5\}$  and  $B = \{3, 4, 6\}$  are  $A$  and  $B$  independent?

## 2 Discrete Random Variables

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## 3 Entropy

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### 3.1 Conditional Entropy

### 3.2 Information Gain

## Part II

# Decision Trees

## Definitions

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Independence, 3  
Probability, 2  
Sample Space, 2

Event, 3  
Subset, 3