

## Learning Goals

---

During this recitation, you will:

- Review common proof techniques
- Build intuition for the connection between induction and recursion
- Review some graph theory

## Proof Techniques

---

**Fundamental proof techniques.** For contradiction and contrapositive, try writing out the truth tables to prove to yourself that these statements are logically equivalent to a direct proof.

- Direct:  $p \implies q$
- Contrapositive:  $\neg q \implies \neg p$
- Contradiction:  $p \wedge \neg q \implies C$
- Induction: To prove  $P(n) \forall n \geq b$ , show  $P(b)$  and  $\forall n \geq b, P(n) \implies P(n+1)$

## Graphs

---

A graph  $G = (V, E)$  consists of a set of vertices (also called nodes)  $V$  together with a set of edges  $E \subseteq V \times V$ , where  $V \times V$  denotes the Cartesian product of  $V$  with itself. We denote an edge from  $u$  to  $v$  as  $(u, v)$ . A graph can be:

- **Undirected:** The edges  $(u, v)$  and  $(v, u)$  are considered the same.
- **Directed:** The edges  $(u, v)$  and  $(v, u)$  are distinct.

In CIS 1210, we do not consider self-loops (i.e., edges of the form  $(v, v)$ ). The **degree** of a vertex  $v$ , denoted  $\deg(v)$ , is the number of edges incident on it. For directed graphs, we distinguish between:

- **In-degree:** Number of edges directed into the vertex.
- **Out-degree:** Number of edges directed out from the vertex.

## Paths and Cycles

A **path** from vertex  $u$  to  $v$  is a sequence:

$$u = v_0, v_1, \dots, v_n = v$$

such that  $(v_i, v_{i+1}) \in E$  for all  $0 \leq i < n$ . The length of a path is the number of edges it contains. A **simple path** contains distinct vertices, meaning  $v_i \neq v_j$  for all  $i \neq j$ .

A **cycle** is a sequence:

$$v_0, v_1, \dots, v_n = v_0$$

where the first and last vertices are the same. Graphs that contain cycles are called **cyclic**, while those that do not are called **acyclic**.

## Connectedness

A graph is said to be:

- **Connected** (for undirected graphs): There is a path between any two vertices.
- **Strongly connected** (for directed graphs): There is a path from  $u$  to  $v$  and from  $v$  to  $u$ .

A (strongly) connected component of a graph is a maximal subgraph that is (strongly) connected.

## Special Types of Graphs

---

A **tree** is an undirected, connected, acyclic graph. Some defining properties of trees include:

- A tree with  $n$  nodes is connected and has exactly  $n - 1$  edges.
- There is a unique path between any pair of vertices.
- Trees are **minimally connected**: removing any edge disconnects the tree.
- Adding any edge to a tree creates a cycle.

A vertex with degree 1 is called a **leaf**. A **forest** is a disjoint collection of trees.

## Binary Trees

A **binary tree** is a tree in which every vertex has at most two children. Typically, the tree is rooted at a specific node, and the terms left and right children are used to describe neighbors. Important properties include:

- The height of a binary tree is the number of edges in the longest path from the root to a leaf.
- The ancestors, descendants, and parents of nodes are defined as expected.

## Random Variables

---

**Definition.** A random variable  $X$  on a sample space  $\Omega$  is a real-valued function that assigns to each sample point  $\omega \in \Omega$  a real number  $X(\omega)$ .

Intuitively you can think of a random variable as mapping an event in your sample space to some value. For example, if  $X$  represents the number of heads after flipping 2 coins,  $X$  maps the event (H, T) to the number 1.

Note, in CIS 1210, you do not need to explicitly define the sample space when dealing with random variables.

**Definition.** Expectation of a random variable denoted  $\mathbb{E}[X]$  is

$$\mathbb{E}[X] = \sum_i i \cdot \Pr[X = i]$$

We can think of the expectation to be what value we “expect”  $X$  to take on after many repeated trials.

**Linearity of Expectation.** For any finite collection of random variables  $X_1, X_2, \dots, X_n$

$$\mathbb{E} \left[ \sum_{i=1}^n X_i \right] = \sum_{i=1}^n \mathbb{E}[X_i]$$