

Lecture 8: Relations and Digraphs

A binary relation R from A to B consists of:

- Set A (Domain)
 - Set B (CoDomain)
 - Subset $\text{graph}(R) \subseteq A \times B$
- $(x, y) \in \text{graph}(R)$ iFF x is "related" to y , $X R Y$ or $R(x, y)$
- $\text{graph}(R)$ is set of all edges defined by the relation.

Can visualize w/ relation diagram.

The relation R^{-1} (inverse R) has domain B , codomain A and $\text{graph}(R^{-1}) := \{(b, a) \in B \times A \mid a R b\}$

$$\rightarrow b R^{-1} a = a R b$$

Range(R) is all the elts w/ arrows coming IN.

Relation Composition: Combining / Chaining Relations

- If $R: A \rightarrow B$ and $T: B \rightarrow C$, then the composition of $R \circ T$, $T \circ R$ is a relation from A to C
- * $T \circ R = T(R(x)) \dots R$ happens 1st.

Functions: Special Case of Binary Relation. $a R b : a R b' \Rightarrow b = b'$

A func from A to B has for all $a \in A$, the set $\{b \in B \mid a R b\}$ has ≤ 1 elt.

↳ ie. Only ≤ 1 one arrow coming out of any $a \in A$.

A function is total if: each $a \in A$ has ≥ 1 outgoing arrow

A function is injective (1-to-1) if: each $b \in B$ has ≤ 1 incoming arrow (F^{-1} exists as func.)

A function is surjective if: each $b \in B$ has ≥ 1 incoming arrow

A function is bijective if: each $a \in A$ has 1 outgoing and each $b \in B$ has 1 incoming

Thm. If f is a total function, then f is bijective iFF it is injective and surjective.

Thm. Bijection from A to $B \Rightarrow |A| = |B|$

Directed Graphs

$G = (V, E)$ has a set V of vertices and set $E \subseteq V \times V$ (edges)

- G is a binary relation on V w/ $E = \text{graph}(G)$
- Can have self-loops, isolated nodes, 2 edges in opp. directions, no edges

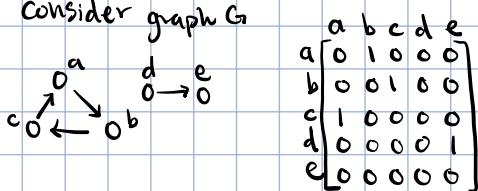
Walks: A sequence $v_1, \dots, v_k \in V$ s.t. $\forall i=1, \dots, k, (v_i, v_{i+1}) \in E$

↳ k vertices are traversed in $k-1$ edges

Path: A walk in which all v_1, \dots, v_k are distinct

Adjacency Matrices

Consider graph G



	a	b	c	d	e
a	0	1	0	0	0
b	0	0	1	0	0
c	1	0	0	0	0
d	0	0	0	0	1
e	0	0	0	1	0

Put 1 in (i, j) iFF $(i, j) \in E$ to get adjacency matrix, A_G

Given $G = (V, E)$ and $n \in \mathbb{N}$, G^n is a relation

$$G^n = (V, E'), E' = \{(u, v) \mid \exists \text{ length } n \text{ walk from } u \text{ to } v\}$$

$\rightarrow u G^n v \text{ iFF } \exists \text{ length } n \text{ walk from } u \text{ to } v.$

Thm. $\forall n \geq 0, (G^n \circ G) = G^{n+1}$

Note on Boolean Matrix Mult. For $n \times n$ matrices $M \in \mathbb{B}^n$, N ,

$$(M * N)_{(i,j)} := \bigvee_{k=1}^n (M[i,k] \wedge N[k,j]), \text{ which returns 1 if either term is 1.}$$

Lemma: $\forall n \geq 0, A_{G^n \circ G} = A_{G^n} * A_G = A_{G^{n+1}}$

Thm. $\forall n \geq 1, A_{G^n} = (A_G)^n$

Connectivity: u connected to v if \exists walk from u to v

Def. $G = (V, E)$, G^* def by $E' = \{(u, v) \mid u \text{ connected to } v\}$

Thm. $G^* = (G^\leq)^{n-1}$ where G^\leq is G w/o all along main diagonal
 $\rightarrow G^\leq := (v \in V \setminus \{(v, v) \mid v \in V\})$

Lecture 9: Partial Orders, Equivalence Relations

Directed Acyclic Graphs (DAGs):

- A directed graph with NO cycles.
 - ex. precedence relations (getting dressed)
- Minimal Elements: Elements with no "in-arrows" (Nothing comes before)
 - * After processing a minimal elt, you might uncover new minimal elts.
 - Can process all minimal elts at once.
- Topological Sort of a DAG is list of nodes such that each node, v , appears earlier in the list than EVERY other node reachable from v .
 - * Every DAG has one or more Topological Sorts
 - ↳ Find by 'Greedy Algorithm': Repeatedly pick minimal elts, remove from graph & add to list.
- Nodes u, v are **comparable** in DAG if u can reach v OR v can reach u . ✗
- A **chain** is a set of nodes s.t. any pair is comparable
- In a chain, a node reachable from all others is the **maximum element** of the chain

Def'n: An **antichain** is a set of nodes s.t. no two nodes are comparable

↳ You can process all minimal nodes in an antichain at the same time!

* Set of all minimal nodes forms an antichain.

1). # rounds required \geq length of any chain

* Longest chain is called **critical path** \Rightarrow # rounds = length of critical path.

2). # rounds \leq length of largest chain/critical path.

↳ if t = length of largest chain, can partition jobs into $\leq t$ antichains

Def'n A partition of set A is a set of nonempty subsets of A s.t. each elt in A is in exactly one block.

Dilworth's Theorem: Every DAG on n nodes has either a chain of size $\geq t$ or antichain of size n/t

Orders:

Linear order: all pairs of diff. elts are comparable. Can tell which comes before/after **ALWAYS**

↳ Has just one topological sort

Binary Relation R on A is:

- Transitive iFF $\forall a, b, c \in A, (aRb \wedge bRc) \Rightarrow (aRc)$
- Reflexive iFF $\forall a \in A, aRa$
- Irreflexive iFF $\forall a \in A, \neg(aRa)$
- Symmetric iFF $\forall a, b \in A, aRb \Rightarrow bRa$
- Asymmetric iFF $\forall a, b \in A, aRb \Rightarrow \neg(bRa)$
- Antisymmetric iFF $\forall a \neq b \in A, aRb \Rightarrow \neg(bRa)$

Strict Partial Order is transitive & irreflexive. $\leftarrow \leftarrow$

\hookrightarrow Thm. R is S.P.O. iFF transitive and asymmetric.

Weak Partial Order is transitive, reflexive, and antisymmetric. $\leq \subseteq$

Defin R is an equivalence relation if it is reflexive, symmetric, and transitive

- Equivalence class of x are all other elements y s.t. xRy

\hookrightarrow Equivalence classes of R on A are blocks of partition of A .

Lecture 10: Simple Graphs: Degrees, Isomorphism, Coloring.

Simple Undirected Graphs:

- Undirected Edges
- At most one edge between vertexes (0 edges ok)
- Good for Symmetric Relationships

$G = (V, E)$ consists of :

- set V of vertices
- subset E of $\{\{u, v\} | u, v \in V, u \neq v\}$ edges

$$\star \{u, v\} = \{v, u\}$$

u is adjacent to v iFF $\{u, v\} \in E$

edge $\{u, v\}$ is incident to $u \& v \dots u \& v$ are endpoints of edge $\{u, v\}$

The degree of $v \in V$, $\deg(v)$, is the # of incident edges to that node

Handshaking Lemma

For every graph $G = (V, E)$, the sum of all degrees is twice the number of edges. $\sum_{v \in V} \deg(v) = 2|E|$

Thm. For $n \geq 1$, the max # edges in an n -node graph is $\frac{n(n-1)}{2}$

Graph Isomorphisms

$G = (V, E)$ and $G' = (V', E')$ are isomorphic if \exists bijection $f: V \rightarrow V'$ that "preserves edges" \leftarrow all edges & non-edges

$$\forall u, v \in V, \{u, v\} \in E \text{ iff } \{f(u), f(v)\} \in E'$$

- Same graph, diff layouts ok
- Same graph, diff node labels ok

P is preserved by isomorphism if for all graphs G, H : $(P(G) \text{ True} \wedge G \text{ is isomorphic to } H) \Rightarrow P(H) \text{ True}$.

Proving Isomorphism: Simply give one example isomorphism

Proving Non-Isomorphism: check #vertices, $\deg(v)$ & v , #edges.

- Find P preserved by isomorphism, then show it doesn't hold for $G \& H$.

Graph Coloring

Idea: Given a graph, assign a color to each vertex s.t. adjacent vertices have diff. colors using min. # colors

Proper Coloring, $C: V \rightarrow \{1, \dots, k\}$ s.t. $\forall \{u, v\} \in E, C(u) \neq C(v)$

Chromatic Number $\chi(G) = \min \# \text{colors needed for proper coloring}$

* Graph containing a triangle has min 3 colors

- Showing a k -coloring shows $\chi(G) \leq k$
- Show how that no $(k-1)$ -coloring works $\Rightarrow \chi(G) = \text{exactly } k$.

C_n is cycle on n vertices: $\chi(C_{2k}) = 2, \chi(C_{2k-1}) = 3$

K_n is clique on n vertices (Max # edges): $\chi(K_n) = n$

4-Color Theorem: Every Planar graph is 4-colorable

Thm: If every vertex in G has degree at most d , then $\chi(G) \leq d+1$

Lecture 11 Simple Graphs: Connectivity, Trees, Bipartite Graphs

For simple graphs, $u \in V$ connected if there is a path between them

As a binary relation, connectivity is:

- Reflexive: $u \in u$ always connected
- Transitive: u connect $v \in v$ connect $w \Rightarrow u$ connect w
- Symmetric: If u connected to v , then v connected to u . **Directed Graphs don't have!**

In simple graphs, connectivity is an equivalence relation.

A graph is connected if it has 1 connected component

↳ Every vertex is connected to every other vertex in a connected component

Trees: G is tree if:

- G is connected and acyclic (does not have any cycles).
- G is connected, but removing any edge in G would disconnect the graph.
- G is acyclic, but adding any edge would create a cycle.
- Every pair of vertices of G is connected by a unique path.
- G is connected and has exactly $n - 1$ edges.
- G is acyclic and has exactly $n - 1$ edges.

Another useful fact:

Theorem 2 (Lemma 12.11.3). Every tree with at least 2 vertices has at least two leaves (i.e., at least two vertices with degree exactly 1).

This is especially useful when doing Induction: start with a tree with $n + 1$ vertices, remove a leaf (we know it has one!), apply the inductive hypothesis, and then reattach the leaf.

Def: A cycle is a sequence v_1, \dots, v_k, v_1 s.t. $k \geq 3$, v_1, \dots, v_k is a path, and $\{v_k, v_1\}$ is an edge.
 $\rightarrow K_3 = C_3 = \text{smallest cycle in undirected graph}$

Def: A cut edge is an edge s.t. removing it disconnects two vertices.

↳ Every edge is cut edge in tree

Lemma: An edge e not cut edge iFF e is on a cycle

Bipartite Graph: Graphs w/two sides

$G = (V, E)$ is bipartite if there are sets $L \subseteq R$ s.t.

- 1). $L \cup R = V$
- 3). $E \subseteq \{\{l, r\} \mid l \in L, r \in R\}$
- 2). $L \cap R = \emptyset$

All edges have one endpoint in L and one endpoint in R .

Thm. G is bipartite iFF G is 2 colorable.

Thm. G is bipartite iFF every cycle has even length.

Thm. Trees are bipartite graphs (bc color each layer alternating colors).

Bipartite Matching:

Goal: Match every $l \in L$ with unique neighbor $r \in R$

edge \leftrightarrow two nodes CAN be matched

Matching: set of $|L|$ edges s.t. every $l \in L$ appears once and every $r \in R$ appears ≤ 1 time.

↳ can have more edges in R than L .

$|L|=|R|$ is perfect matching

Hall's Theorem: There is matching iFF there are no bottlenecks ($\forall S \subseteq L, |N(S)| \geq |S|$)
where $N(S) = \{r \in R \mid \exists l \in S \text{ s.t. } \{l, r\} \in E\}$

Lecture 12: Evaluating Sums

For a summation, a closed-form solution is an algebraic equation with the same value as the summation

Perturbation Method:

Combine a sum and a perturbation of that sum to get something useful.

$$\text{ex. } S = \sum_{i=0}^{n-1} x^i \quad S = 1 + x + x^2 + \dots + x^{n-1}$$
$$XS = x + x^2 + \dots + x^{n-1} + x^n \Rightarrow S - XS = 1 - x^n \quad S = \frac{1 - x^n}{1 - x}$$

Derivative Method: Use derivative of known closed forms to find new sums

$$\text{ex. } S = \sum_{i=0}^n i \cdot x^i$$

Notice S looks like $\sum_{i=0}^n i x^{i-1} = \frac{d}{dx} \left(\sum_{i=0}^n x^i \right)$, where we know $\sum_{i=0}^n x^i = \frac{1 - x^{n+1}}{1 - x}$

$$\rightarrow \sum_{i=0}^n i x^{i-1} = \frac{1 - (n+1)x^n + nx^{n+1}}{(1-x)^2} \quad \text{... multiply both sides by } X \text{ to get what we want!}$$

Smart Grouping: Identify groupings to make your life easier!

$$\text{ex. } S = \sum_{i=1}^n i : \quad S = 1 + 2 + \dots + n \quad \text{AND} \quad S = n + n - 1 + \dots + 1$$
$$\text{so } 2S = n(n+1) \text{ and } S = \frac{n(n+1)}{2}$$

Approximating Sums with No Closed Form:

Method of Integration Bounds:

- if f is positive & increasing: $f(1) + \int_1^n f(x) dx \leq \sum_{i=1}^n f(i) \leq f(n) + \int_n^{\infty} f(x) dx$
- * flipped if decreasing.
- $f(x)$ is what is inside the summation btw.

Lecture 13: Asymptotics

Stirling's Formula: $n! \sim \left(\frac{n}{e}\right)^n \sqrt{2\pi n}$

<u>Notation</u>	<u>Test / Rule</u>	
$f \sim g$	$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$	$f \in \omega(g) \Rightarrow f \notin O(g)$
$f \in O(g)$	$\lim_{n \rightarrow \infty} \frac{ f(n) }{g(n)} \in \mathbb{R}$ $\exists c \exists M. \forall x > M. f(x) \leq c g(x)$	$f \in o(g) \Rightarrow f \notin \Omega(g)$
$f \in o(g)$	$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$	\leq
$f \in \Omega(g)$	$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} \in (0, \infty]$ when $g \in O(f)$	$<$
$f \in \omega(g)$	$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$ when $g \in o(f)$	\geq
$f \in \Theta(g)$	$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} \in (0, \infty)$ $f \in O(g)$ and $f \in \Omega(g)$	$>$

Lecture 14: Recurrences

Recursion is like induction on algorithms!

$A(x)$: if x small ... some stuff (base case)
 Else: ... some stuff
 call $A(x')$ where x' is "smaller" than x . (recursive step)
 ... stuff

Counting # of moves an algorithm takes:

- Let $T(n)$ be # moves for n elements.
- Find # moves for every step in the alg. (May be written as $T(\text{something})$).
- Combine to find $T(n)$, the open form solution.

Find closed form of $T(n)$ by:

1). Guess and Verify

- Plug in small values of n
- Guess pattern
- Prove guess by induction

2). Plug & Chug (Expansion, Iteration, Brute Force)

- Plug in expansions of recursive terms where possible. Simplify
- Repeat until you can guess a pattern
- Prove guess by induction

↳ Guess may still have recursive term. Find params s.t. that $T(\text{something})$ terms disappears. See lec ex.

Divide and Conquer Recurrence:

A D.C recurrence given by $T(x) = g(x) + \sum_{i=1}^K a_i T(b_i x)$ for $x > x_0$ that satisfies:

- 1). For all i , $a_i > 0$ and $0 < b_i < 1$
- 2). g is differentiable and polynomially bounded
- 3). X_0 is "big enough"

Thm Akra-Bazzi

If $T(n)$ has D.C recurrence, and

$$P \text{ s.t. } \sum_{i=1}^K a_i b_i^P = 1 \quad \text{Increasing } P \text{ decreases sum bc } b < 1 \quad \star$$

Then, $T(n) \in \Theta\left(n^P + n^P \int_1^n \frac{g(x)}{x^{P+1}} dx\right)$

When P is not easy to find: find what P must be less than s.t. $\sum_{i=1}^K a_i b_i^P < 1$

↳ Then, execute Akra-Bazzi and find $T(n)$ with P still in exponents
Then, find dominating term given the bound on P we found.

Thm. If $g(x) \in \Theta(x^t)$ and $t \geq 0$ and $\sum_{i=1}^K a_i b_i^t < 1$

Then, $T(x) \in \Theta(g(x))$

Master Theorem (Simpler cases of Akra-Bazzi for $K=1$) Here, $p = -\log_b(a_1)$

case 1: If $g(x) \in O(x^c)$ $c < p$
then $T(x) \in \Theta(x^p)$

case 2: If $g(x) \in \Theta(x^p)$
then $T(x) \in \Theta(x^p \log x)$

case 3: If $g(x) \in \Omega(x^c)$ $c > p$
then $T(x) \in \Theta(g(x))$

RANDOM!

- Directed graphs have self-loops, which are cycles.
- Remember prove both directions for iff

Some Common Sums:

$$\text{For } |x| < 1, \sum_{i=0}^{\infty} x^i = \frac{1}{1-x}. \text{ Diverges if } |x| \geq 1$$

$$\sum_{k=1}^n k = \frac{n(n+1)}{2}$$

$$\sum k^2 = \frac{n(n+1)(2n+1)}{6}$$

$$\sum k^3 = \frac{n^2(n+1)^2}{4}$$

$$\text{Harmonic Series } H_n = \sum_{i=1}^n \frac{1}{i}$$

$$\sum_{i=1}^{\infty} i x^{i-1} = \frac{1}{(1-x)^2}$$

$$\sum_{i=0}^n ar^i = \frac{ar^{n+1} - a}{r-1}$$

Apply log to a product turns it into a sum:

$$\ln(n!) = \ln(1) + \ln(2) + \dots + \ln(n) = \sum_{i=1}^n \ln(i).$$

Proofs By Contradiction when you need to prove something can't happen.

Quiz 2 Review 6.042

Lecture 8 - Relations and Digraphs (Directed Graphs)

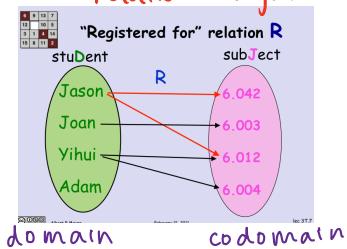
Relations + Functions

- Binary Relations - associates elements of domain with elements of codomain
 - Notation: $J \in R$ is "registered for" (R) 6.042
 - Jason $\in R$ 6.042 infix "less than" relates pairs of #l's
 - $R(Jason, 6.042)$ prefix "subset" relates pairs of sets.
 - $(Jason, 6.042) \in R$
 - $(Jason, 6.042) \in \text{graph}(R)$ subset graph(R) $\subseteq A \times B$
↳ pairs of elements $a \in A, b \in B$ (a, b).
- $R(X) ::=$ everything R relates to things in X ,
endpoints of arrows from points in X
 $\{j \in J \mid \exists d \in X. d R j\}$
 arrow from $X \rightarrow j$

Left

Right

Relation Diagram



ex. $R(\text{Jason})$,
 $R(\{\text{Jason}, \text{Yihui}\})$

ex. $R^{-1}(6.012)$

- R^{-1} (inverse image): $d R j \iff j R^{-1} d$, ↪
 - $D \subseteq R^{-1}(J) \Rightarrow$ every student is registered

Domain: Codomain Swapped and $\text{graph}(R^{-1}) := \{(b, a) \in B \times A \mid a R b\}$

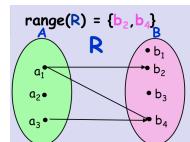
- Composition: $R \circ V(X) ::= R(V(X))$ * V happens first, then R
 - $R \circ V ::=$ "prof has advisee registered for"
 - $p(R \circ V) j ::=$ prof p has an advisee in subject j
 - Ex. ARM $(R \circ V)$ 6.012 b/c ARM V Yihui and Yihui R 6.012

Combining
relations

Set operations on relations

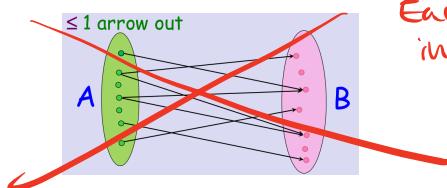
- $T \cap (R \circ V) = \emptyset$, Professors should not teach advisees
 $R \circ V \subseteq \overline{T}$

- Range(R) ::= elements with arrows coming in
all elements in
codomain connected
to domain.



- Function: relation which associates each element, a , of A with at most 1 element of B (1 arrow out)

$[a F b \text{ AND } a F b'] \text{ IMPLIES } b = b'$



Each element in A points to at most one element in B

↳ One x value in a func. can point to at most one y value.

Relational Mapping Properties

OUT ARROWS

- function: ≤ 1 arrow out
- total relation: ≥ 1 arrow out
 $\rightarrow R$ is total IFF $A = R^{-1}(B)$
- total and function: exactly 1 arrow out

ex. $\frac{1}{x-y}$ is not total, $g(r, r)$ DNE

IN ARROWS

- surjection: ≥ 1 arrow in (see if TR is possible for outputs)
 $\rightarrow R$ is a surjection IFF $R(A) = B$
- injection: ≤ 1 arrow in (check if multiple inputs \rightarrow 1 output)
- bijection: exactly 1 arrow out, exactly 1 arrow in
 \rightarrow implies $|A| = |B|$
 \rightarrow must be: surjection, injection, total, function

If f is total, then f is bijection IFF f is injective and surjective.

Directed Graphs (Digraphs)

- Digraphs:
 - a set, V , of vertices
 - a set, $E \subseteq V \times V$ of directed edges
 - $(v, w) \in E$
 - notation: $v \rightarrow w$
 - a digraph with vertices V is the same as a binary relation on V
- Walks: follow successive edges K vertices are traversed in K-1 edges
- Paths: walk thru vertices without repeating vertex... a more specific walk
- Adjacency Matrix

Given $G = (V, E)$ and $n \in \mathbb{N}$, G^n is a relation

Connected vertices $G^n = (V, E')$, $E' = \{(u, v) \mid \exists \text{ length } n \text{ walk from } u \text{ to } v\}$

- Lemma: the shortest walk between two vertices is a path!
- length n walk relation
 - $v G^n m$ IFF \exists length n walk from v to m ,
 - G^n is the length n walk relation for G
 - Lemma: $G^m \circ G^n = G^{m+n}$
- Matrices & Composition: A_G := adjacency matrix for G
 - Lemma: $A_{G \circ H} = A_H \odot A_G$
 - $\star ? ?$ \odot boolean matrix prod (using or)
- Walk Relation G^* : G^* is walk relation of G $u G^* v$ IFF \exists walk u to v (u is connected to v)

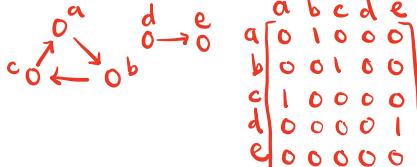
$G^*, G^+ ?$

Lemma: $\forall n \geq 0, A_{G \circ G} = A_{G^n} * A_G = A_{G^{n+1}}$

Boolean Matrix Mult: $(M * N)(i,j) := \bigvee_{k=1}^n (M[i,k] \wedge N[k,j])$

returns 1 if at least one of our entries $M[i,k] \wedge N[k,j]$ is 1.

Consider graph G



	a	b	c	d	e
a	0	1	0	0	0
b	0	0	1	0	0
c	1	0	0	0	0
d	0	0	0	0	1
e	0	0	0	0	0

$\forall n \geq 1, A_{G^n} = (A_G)^n$

Put 1 in (i,j) IFF $(i,j) \in E$ to get adjacency matrix, A_G

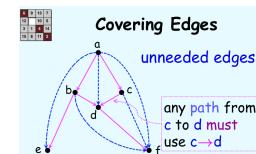
Connectivity: u is connected to v if \exists walk from u to v .

Thm. $G^* = (G^\leq)^{n-1}$ where G^\leq is G w/ 1's all along main diagonal ?

Lecture 9 - Partial Orders, Equivalence Relations

Directed Acyclic Graphs (DAGs)

- indirect prerequisite: there is a sequence of prereqs from u to v
→ aka there is a positive length walk from u to v ($u D^+ v$)
- closed walk: start + end on same vertex
- cycle: closed walk whose only repeat vertex is its start
→ 1 vertex = length 0 cycle
- Lemma: shortest positive length closed walk from v is positive length cycle from v
→ proof: like walk / path
- directed acyclic graph (DAG): has no positive length cycle
(or closed walk)
→ ex. prerequisite graph, $\subset \{a, b\} \rightarrow \{a, b, d\}$
→ covering edge: needed edge



DAG's & scheduling

- a minimal subject: no prerequisites
→ nothing $\rightarrow d$, $\forall x \in A$ s.t. $\forall y \neq x, y \not\leq x$
- minimum subject: earliest of all (an indirect prereq)
- "greedy": take max subjects each term
- antichain: set of subjects with no indirect prereqs, "incomparable"
→ u is incomparable to v iff no path from u to v and no path from v to u
- topological sort: take classes by layer, every finite poset has a topological sort
- chain: sequence of subjects that must be taken in order
→ max chain length \leq # terms to graduate

if neither $a \leq b$
nor $b \leq a$
 \Rightarrow

Partial Orders

- strict partial orders: transitive + asymmetric
→ ex: • C on sets
 - "indirect prerequisite"
 - less than, $<$, on real #s
- R is a SPO IFF $R = D^+$ for some DAG
- linear^{path-total} orders: given any 2 elements, 1 will be "bigger than" the other
→ ex: $<$ or \leq on reals
→ no incomparable elements — $x R y$ or $y R x$
→ whole partial order is a chain
- topological sort: turns a partial order into a linear order... in a way that is consistent w/ partial order
- weak partial order: same as strict partial order R , except that
 - poset $a R a$ always holds
 - ex. \leq, \subseteq
 - transitive, antisymmetric, reflexive (all elements)
 - Thm: R is a WPO iff $R = D^*$ for some DAG D
- Notation: $a \leq b$ — a comes before b in partial order
 - poset: partially ordered set (A, \leq)
 - same as a DAG (except self-loops)

Summary of Relational Properties (10.11 in tx+bk) * see Rec 9 notes

* Relation $R: A \rightarrow A$ is the same as a digraph with vertices A

" R is ____ when"

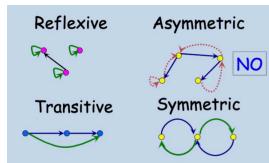
- Reflexivity: $\forall x \in A. x R x$ (every vertex has self loop)
- Irreflexivity: $\text{NOT } [\exists x \in A. x R x]$ (no self-loops in R)
- Symmetry: $\forall x, y \in A. x R y \text{ IMPLIES } y R x$ (if edge $x \rightarrow y$, then $\exists y \rightarrow x$ in R)
- Asymmetry: $\forall x, y \in A. x R y \text{ IMPLIES NOT}(y R x)$ (at most 1 directed edge, no self-loops)
- Antisymmetry: $\forall x \neq y \in A. x R y \text{ IMPLIES NOT}(y R x)$ (" ", may be self-loops)
or $\forall x, y \in A. (x R y \text{ AND } y R x) \text{ IMPLIES } x = y$
- Transitivity: $\forall x, y, z \in A. (x R y \text{ AND } y R z) \text{ IMPLIES } x R z$ (if path from $u \rightarrow v$, edge from $u \rightarrow v$)
- Linear: $\forall x \neq y \in A. (x R y \text{ OR } y R x)$ (given any 2 vertices, \exists some edge)

Partial Orders:

- Strict Partial Order: R is a strict partial order IFF
 - R is transitive and irreflexive IFF
 - R is transitive and asymmetric IFF
 - R is the positive length walk relation of a DAG
 - Weak Partial Order: R is a weak partial order IFF
 - R is transitive and anti-symmetric and reflexive IFF
 - R is the walk relation of a DAG
 - Equivalence Relation: R is an equivalence relation IFF
 - R is reflexive, symmetric, transitive IFF
 - R equals the in-the-same-block-relation for some partition of domain(R)
- ex. \subset on sets
"indirect prerequisite"
 $<$ on real #s
- ex. \leq, \subseteq
- ex. $=, \equiv$, same size, same color

Equivalence Relations

- two-way walks: $u G^* v$ AND $v G^* u$, u and v are strongly connected
- equivalence relations: transitive, symmetric, reflexive
 - R is equiv. rel. IFF R is strongly connected relation of digraph
 - ex. $=, \equiv$, same size/color
- Graphical Properties of Relations
- Representing equivalences
 - for total function $f: A \rightarrow B$, define relation \equiv_f on A : $a \equiv_f a'$ iff $f(a) = f(a')$
 - Thm: R on set A is equiv. relation IFF R is \equiv_f for some $f: A \rightarrow B$
 - $\equiv_{(\text{mod } n)}$ is \equiv_f (where $f(k) := \text{rem}(k, n)$)
 - for partition Π of A , def. relation \equiv_Π on A : $a \equiv_\Pi a'$ IFF a, a' are in same block of Π
 - Thm: R on A is an equiv. rel. IFF R is \equiv_Π for some partition Π of A



Posets (Partially Ordered Sets) - relation that's transitive, antisymmetric, reflexive

- Hasse diagram - a way of representing a poset (A, \leq) as a DAG
- chain - a subset $C \subseteq S$ in poset (S, \leq) . $\forall x, y \in C$, either $x \leq y$ or $y \leq x$
- antichain - a subset $A \subseteq S$. $\forall x, y \in A$, $x \neq y$, neither $x \leq y$ nor $y \leq x$
- topological sort - total order of all the elements such that if $a_i \leq a_j$ in the partial order, then a_i precedes a_j in the total order

Lecture 10: Simple Graphs - Isomorphism, Coloring

Simple Graphs: Degrees

- simple graph G - consists of
 - nonempty set V of vertices
 - a set E of edges such that each edge has 2 endpoints in V
 - undirected edges
- vertex degree - # of incident edges (that touch it)
- Handshaking Lemma - $2|E| = \sum_{v \in V} \deg(v)$ (sum of degrees = twice # edges)

Simple Graphs: Isomorphism

- isomorphic: \exists bijection $f: V_1 \rightarrow V_2$ with $u-v$ in E_1 IFF $f(u)-f(v)$ in E_2
 - aka there is an edge-preserving matching of their vertices (that is bijective)
 - NOT isomorphic if any properties differ:
 - # nodes
 - # edges
 - degree distributions
 - length of any cycles that exist

Simple Graphs: Coloring

- planar graph: if it is possible to find a planar drawing of that graph (a drawing where the edge-lines don't intersect each other)
- coloring: assignment of colors to each vertex in a graph G so no edge has the same coloring on both ends
 - Any planar map is 4-colorable.
 - Note: 7-colorable \neq not 6-colorable
- chromatic number $\chi(G)$: minimum # of colors to validly color G
 - G is K -colorable iff $\chi(G) \leq K$
 - for a complete graph K_n , $\chi(K_n) = n$
 - $\chi(C_{\text{even}}) = 2$
 - $\chi(C_{\text{odd}}) = 3$
 - all degrees $\leq K$ implies $\chi(G) \leq K+1$

Graph Proof: Inductive Step

- ① Start with $G(n+1)$ that satisfies desired constraints
- ② Remove some vertex (or edge) to get $G(n)$
- ③ Prove $G(n)$ satisfies desired constraints
- ④ Apply IH to $G(n)$
- ⑤ Add vertex back and show $P(n+1)$ holds

Solution. *Proof.* By induction on the number of vertices n . The induction hypothesis is

$$P(n) := \text{all } n\text{-vertex graphs } G \text{ with width one are forests.}$$

(Rec 11)

Base case: ($n = 1$). A graph with one vertex is acyclic and therefore is a forest.

Induction step. Assume that $P(n)$ is true for some $n \geq 1$ and let G be an $(n+1)$ -vertex graph with width one. We need only show that G is acyclic.

The vertices of G can be listed with each vertex adjacent to at most one vertex earlier in the list. Let v be the last vertex in the list. Since *all* the vertices adjacent to v appear earlier in the list, it follows that $\deg(v) \leq 1$.

Now removing a vertex won't increase width, so $G \setminus v$ (G without v) still has width one, so it is acyclic by Induction Hypothesis. But no degree-one vertex is in a cycle, so adding v back to $G \setminus v$ will not create a cycle. Hence G is acyclic, as claimed.

This proves $P(n+1)$ and completes the induction step.

Lecture 11: Bipartite Matching, Trees

Simple Graphs: Connectivity

- A graph is connected iff all its vertices are connected to each other
 - has only 1 connected component
- 2 vertices are connected iff there is a path btwn them
- Connected component: $\underbrace{E^*(v)}_{\substack{\text{walk} \\ \text{relation}}}$

Trees

- Tree definitions: (more below)
 - A connected graph with no cycles
 - Every edge is a cut edge
 - A connected graph that is edge minimal
 - Connected graph with n vertices and $n-1$ edges
 - An edge-maximal acyclic graph
 - Graph with a unique path btwn any 2 vertices
- Cut edge - removing the edge disconnects 2 vertices
 - Lemma: an edge is not a cut edge iff it is on a cycle
- Leaf: vertex of degree at most 1
- Trees:

Theorem 1 (Theorems 12.11.4 and 12.11.6). *For an undirected graph G with $n = |V(G)|$ vertices, the following statements are all equivalent to G being a tree:*

- G is connected and acyclic (does not have any cycles).
- G is connected, but removing any edge in G would disconnect the graph.
- G is acyclic, but adding any edge would create a cycle.
- Every pair of vertices of G is connected by a unique path.
- G is connected and has exactly $n - 1$ edges.
- G is acyclic and has exactly $n - 1$ edges.

Another useful fact:

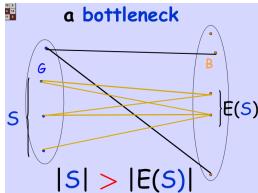
Theorem 2 (Lemma 12.11.3). *Every tree with at least 2 vertices has at least two leaves (i.e., at least two vertices with degree exactly 1).*

This is especially useful when doing Induction: start with a tree with $n + 1$ vertices, remove a leaf (we know it has one!), apply the inductive hypothesis, and then reattach the leaf.

- Tree coloring:
 - $\chi(G) = 2$ (all cycles, if any, are even len)

Bipartite Matching

- **Bipartite graph:** vertices can be partitioned into 2 blocks, $L(G)$ and $R(G)$ such that every edge has 1 endpoint in $L(G)$ and the other in $R(G)$
 - graph is bipartite iff it is 2-colorable
- **bottleneck:** a set S of girls without enough boys, $|S| > |E(S)|$



Hall's Theorem

- **bipartite match:** a total injective function $L(H) \rightarrow R(H)$
 - matching: any pairing of some subset of L with R such that no element is paired w/ > 1 element
 - perfect: includes all elements
- **Hall's Theorem:** If there are no bottlenecks, then there is a match, aka

If $|S| \leq |E(S)|$ for all sets $S \subseteq L(H)$,

 then there is a match.
Hall's condition
- **Thm 12.5.6.** - If G is a degree-constrained bipartite graph, then there is a matching that covers $L(G)$.
 - degree-constrained: if $\min \text{ degree}(L) \geq \max \text{ degree}(R)$ [Rec 11]
- **Proof of Hall's Thm**
 - Case 1) $|S| = |E(S)|$
 - Lemma. If S a set of girls with $|S| = |E(S)|$, then no bottlenecks btwn S and $\bar{E}(S)$ either
 - so, no bottlenecks in $(S, E(S))$ and none in $(\bar{S}, \bar{E}(\bar{S}))$
 - Case 2) $|S| < |E(S)|$
 - match g with b
 - removing b still leaves $|S| \leq |E(S)|$, so no bottlenecks.
 - by induction, can match remaining $g + b$

2 Hall's Matching Theorem

We use the notation $G = (L, R, E)$ for a bipartite graph with vertices $V = L \cup R$, where L and R are disjoint sets representing the “left” vertices and “right” vertices, respectively, and where every edge has exactly one endpoint in L and one endpoint in R .

A **matching** in G is a subset of edges $M \subseteq E$ so that every vertex in G lies on *at most* one edge from M .

If M is a matching and every vertex $v \in L$ belongs to an edge in M , we say that M **covers L** . (Define **covers R** similarly.) If M covers both L and R , we call M a **perfect matching** (note that this requires $|L| = |R|!$).

A subset $S \subseteq L$ is called a **bottleneck** in G if it connects to strictly fewer than $|S|$ neighbors in R , via edges in E .

We have the **quick observation** that having a bottleneck immediately shows that there can't be a matching that covers L : if $S \subseteq L$ is a bottleneck, then the vertices in S collectively have fewer than $|S|$ options to choose from, so someone from S will always be left out. So avoiding bottlenecks is a **necessary condition** for having a matching that covers L .

The **more surprising** fact is that this condition is **also sufficient**:

Theorem 3 (Hall's Matching Theorem (Theorem 12.5.4)). *A bipartite graph $G = (L, R, E)$ has a matching that covers L if and only if no subset $S \subseteq L$ is a bottleneck.*

Lecture 12: Sums

Example: Pricing an annuity

def: A n -year \$m-payment annuity pays \$m at start of each year for n years

Assumption: fixed interest rate p

- \$1 today = \$\$(1+p) in 1 year
- $\frac{1}{1+p}$ today = \$1 in 1yr
- $\frac{1}{(1+p)^2}$ today = \$1 in 2 yrs
⋮

current value

$$\begin{aligned} \$m \\ \$m/(1+p) \end{aligned}$$

payments

$$\begin{aligned} \$m \text{ now} \\ \$m \text{ in 1 yr} \end{aligned}$$

⋮

$$\$m/(1+p)^{n-1}$$

\$m in $n-1$ yrs

$$V = \sum_{i=0}^{n-1} m/(1+p)^{n-i} = m \sum_{i=0}^{n-1} x^i; x = \frac{1}{1+p}$$

$$\boxed{\sum_{i=0}^{n-1} x^i = \frac{1-x^n}{1-x}}$$

Perturbation: line up terms, get cancellation

$$\begin{aligned} S &= 1+x+x^2+\dots+x^{n-1} \\ -xS &= x+x^2+\dots+x^{n-1}+x^n \\ (1-x)S &= 1-x^n \Rightarrow S = \frac{1-x^n}{1-x} \end{aligned}$$

Claim: If $n=\infty$, $V=m\left(\frac{1+p}{p}\right)$

Proof: $\lim_{n \rightarrow \infty} \left(\frac{1}{1+p}\right)^n = 0 \Rightarrow \lim_{n \rightarrow \infty} V = \frac{m}{1-\frac{1}{1+p}} = \left(\frac{1+p}{p}\right)m \quad \square$

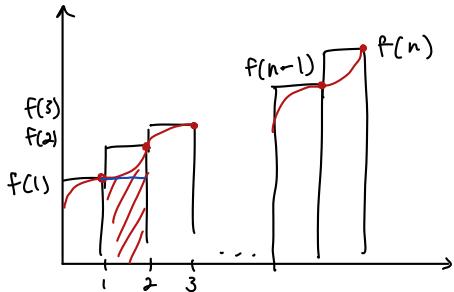
Corollary: If $|x|<1$, $\sum_{i=0}^{\infty} x^i = \frac{1}{1-x}$ e.g. $1 + \frac{1}{2} + \frac{1}{4} + \dots = \frac{1}{1-\frac{1}{2}} = 2$
 $1 + \frac{1}{3} + \frac{1}{9} + \dots = \frac{1}{1-\frac{1}{3}} = \frac{3}{2}$

Derivative: $\sum_{i=0}^n x^i = \frac{1-x^{n+1}}{1-x}$

$$\int (\sum x^i) = \sum \frac{x^{i+1}}{i+1}$$

$$\begin{aligned} x \left(\sum_{i=0}^{n-1} i x^{i-1} = \frac{-(n+1)x^n(1-x)+(1-x^{n+1})}{(1-x)^2} \right) \\ \sum_{i=1}^{n-1} i x^i = f(x) \cdot x \end{aligned}$$

Integration Bounds: If f is positive and increasing



$$I = \int_1^n f(x) dx$$

If f is weakly increasing + $S := \sum_{i=1}^n f(i)$,

$$I + f(1) \leq S \leq I + f(n)$$

If f is weakly decreasing, then

$$I + f(n) \leq S \leq I + f(1)$$

Summation Rules

$$1. \sum_{i=1}^n c = cn$$

$$2. \sum_{i=1}^n c \cdot a_i = c \sum_{i=1}^n a_i$$

$$3. \sum_{i=1}^n (a_i + b_i) = \sum_{i=1}^n a_i + \sum_{i=1}^n b_i$$

$$4. \sum_{i=1}^n i = \frac{n(n+1)}{2}$$

$$5. \sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6}$$

$$6. \text{geometric sum } \sum_{i=0}^{n-1} x^i = \frac{1-x^n}{1-x} \quad \sum_{k=0}^{\infty} ar^k = \frac{a}{1-r} \quad \sum_{k=j}^n 2^k = 2^{n+1} - 2^j$$

* review double sums

Lecture 13: Asymptotics

Stirling's Formula: $n! = \left(\frac{n}{e}\right)^n \sqrt{2\pi n} e^{\Sigma(n)}$ where $\frac{1}{2n+1} \leq \Sigma(n) \leq \frac{1}{2n}$

$$100! \geq \left(\frac{100}{e}\right)^{100} \sqrt{200\pi} e^{\frac{1}{1201}}$$

$$100! \leq \left(\frac{100}{e}\right)^{100} \sqrt{200\pi} e^{\frac{1}{1200}}$$

$$n! \sim \left(\frac{n}{e}\right) \sqrt{2\pi n}$$

Asymptotic Notation:

Name	Notation	Explanation	Summary
"tilde"	$f(x) \sim g(x)$	$\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = 1$	
"O" / "Big O"	$f(x) = O(g(x))$	$\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} < \infty$	\leq
"omega"	$f(x) = \Omega(g(x))$	$\lim_{x \rightarrow \infty} \left \frac{f(x)}{g(x)} \right > 0$	\geq
"theta"	$f(x) = \Theta(g(x))$	$\lim_{x \rightarrow \infty} \left \frac{f(x)}{g(x)} \right > 0, < \infty$	$=$ if Θ , then O and Ω
"little o"	$f(x) = o(g(x))$	$\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} < \infty$	$<$
"little omega"	$f(x) = \omega(g(x))$	$\lim_{x \rightarrow \infty} \left \frac{f(x)}{g(x)} \right > 0$	$>$

Perturbation Method (P13)

$$S = \sum_{k=0}^{\infty} \frac{F_k}{2^k}$$

$$S = F_0 + \frac{F_1}{2} + \frac{F_2}{2^2} \dots + \frac{F_n}{2^n}$$

$$\frac{S}{2} = \frac{F_0}{2} + \frac{F_1}{4} + \dots + \frac{F_n}{2^{n+1}}$$

$$S + \frac{S}{2} = F_0 + \sum_{i=0}^{\infty} \frac{F_i + F_{i+1}}{2^{i+1}} = F_0 + \sum_{i=0}^{\infty} \frac{F_{i+2}}{2^{i+1}} = F_0 + 2 \left(S - \frac{F_0}{2^0} - \frac{F_1}{2^1} \right)$$

$$\sum \frac{F_{i+2}}{2^{i+1}} \Rightarrow \frac{F_2}{2} + \frac{F_3}{2^2}$$

$$f \in \omega(g) \Rightarrow f \notin O(g)$$

$$f \in o(g) \Rightarrow f \notin \Omega(g)$$

Lecture 13:

Lecture 14: Divide + Conquer Recurrences

(Review Rec)

- Def: $T_n = \min \# \text{ moves needed for } n \text{ discs}$

Guess + Verify (aka substitution)

$$\text{Guess: } T_n = 2^n - 1$$

Verify: by induction

$$P(n): T_n = 2^n - 1$$

$$\underline{\text{Base Case: }} n=1, T_1 = 1 = 2^1 - 1 = 1 \quad \checkmark$$

$$\underline{\text{Ind Step: }} \text{Assume } T_n = 2^n - 1 \text{ to prove } T_{n+1} = 2^{n+1} - 1$$

$$T_{n+1} = 2T_n + 1 = 2(2^n - 1) + 1 = 2^{n+1} - 2 + 1 = 2^{n+1} - 1 \quad \checkmark$$

Merge Sort

To sort $n > 1$ items x_1, x_2, \dots, x_n ($n = \text{power of 2}$)

1. sort $x_1, x_2, \dots, x_{\frac{n}{2}}$ & $x_{\frac{n}{2}+1}, x_{\frac{n}{2}+2}, \dots, x_n$ recursively

2. merge

Def. $T(n) = \# \text{ comparisons used to sort } n \text{ numbers}$

Merging takes $n-1$ comparisons (worst case)

recursive sorting takes $2T(\frac{n}{2})$ comparisons

$$\Rightarrow T(n) = n-1 + 2T(\frac{n}{2})$$

Plug & Chug (aka Expansion, Iteration, Brute force)

$$\begin{aligned} T(n) &= n-1 + 2T(\frac{n}{2}) \\ &= n-1 + 2\left(\frac{n}{2}-1 + 2T(\frac{n}{4})\right) = n-1 + n-2 + 4T(\frac{n}{4}) \\ &= n-1 + n-2 + 4\left(\frac{n}{4}-1 + 2T(\frac{n}{8})\right) = n-1 + n-2 + n-4 + 8T(\frac{n}{8}) \\ &= n-1 + n-2 \dots + n-2^{\frac{n}{2}-1} + 2^{\frac{n}{2}}T(\frac{n}{2}) \quad i = \log n \\ &= n-1 + n-2 \dots + n-2^{\log n-1} + 2^{\log n}T(\frac{n}{2^{\log n}}) \\ &= \sum_{i=0}^{\log n-1} (n-2^i) = n\log n - (2^{\log n} - 1) = n\log n - n + 1 \end{aligned}$$

$$\left\{ \begin{array}{l} T(n) = 2T(n-1) + 1 \Rightarrow T(n) \sim 2^n \\ T(n) = 2T(\frac{n}{2}) + n-1 \Rightarrow T(n) \sim n\log n \\ S(n) = 2S(\frac{n}{2}) + 1 \Rightarrow S(n) \sim n \end{array} \right. \quad \text{divide + conquer}$$

Divide + Conquer Recurrences have the form

$$T(x) = a_1 T(b_1 x + \sum_i(x)) + a_2 T(b_2 x + \sum_2(x)) + \dots + a_k T(b_k x + \sum_k(x)) + g(x) \text{ for } x \geq x_0$$

$$T(x) = \Theta(1) \text{ for } x \leq x_0$$

where: $a_i > 0, 0 < b_i < 1, k \text{ fixed}$

$$|\sum_i(x)| \leq O\left(\frac{x}{\ln^2 x}\right)$$

$$|g(x)| \leq x^c \text{ for some } c$$

Akra-Bazzi Theorem

Theorem 1 (Akra-Bazzi, strong form). Suppose that:

$$T(x) = \begin{cases} \text{is defined} & \text{for } 0 \leq x \leq x_0 \\ \sum_{i=1}^k a_i T(b_i x + h_i(x)) + g(x) & \text{for } x > x_0 \end{cases}$$

where:

- a_1, \dots, a_k are positive constants
- b_1, \dots, b_k are constants strictly between 0 and 1
- x_0 is “large enough” in a technical sense we leave unspecified
- $|g'(x)| = O(x^c)$ for some $c \in \mathbb{N}$
- $|h_i(x)| = O(x/\log^2 x)$

Then:

$$T(x) = \Theta\left(x^p \left(1 + \int_1^x \frac{g(u)}{u^{p+1}} du\right)\right)$$

where p satisfies the equation $\sum_{i=1}^k a_i b_i^p = 1$.

example:

The only difference between the strong and weak forms of Akra-Bazzi is the appearance of this $h_i(x)$ term in the recurrence, where $h_i(x)$ represents a small change in the size of the subproblems ($O(x/\log^2 x)$). Notice that, despite the change in the recurrence, the solution $T(x)$ remains the same in both the strong and weak forms, with no dependence on $h_i(x)$! In algorithmic terms, this means that *small* changes in the size of subproblems have no impact on the asymptotic running time. **Example:** Let’s compare the Θ bounds for the following

divide-and-conquer recurrences.

$$T_a(n) = 3T\left(\frac{n}{3}\right) + n \quad T_b(n) = 3T\left(\left\lfloor \frac{n}{3} \right\rfloor\right) + n$$

For $T_a(n)$ we have $a_1 = 3$, $b_1 = 1/3$, $g(n) = n$, $p = 1$. For $T_b(n)$ we have the same parameters as for $T_a(n)$, plus $h_1(n) = \lfloor n/3 \rfloor - n/3$. Using the strong Akra-Bazzi form, the $h_1(n)$ falls out of the equation:

$$T_a(n) = T_b(n) = \Theta\left(n\left(1 + \int_1^n \frac{u}{u^2} du\right)\right) = \Theta(n \log n).$$

The addition of the ceiling operator changes the value of $n/3$ by at most 1, which is easily $O(n/\log^2 n)$. So floor and ceiling operators have no impact on the asymptotic solution to a recurrence.