Outline for Today

- Cuckoo Hashing
 - A simple, fast hashing system with worstcase efficient lookups.
- The Erdős-Rényi Model
 - Randomly-generated graphs and their properties.
- Variants on Cuckoo Hashing
 - Making a good idea even better.

Preliminaries: Hash Tables

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 - **Open addressing:** Allow elements to overflow out of their target bucket and into other spaces. (For example, linear probing hashing.)
 - **Perfect hashing:** Do something clever with multiple hash functions to eliminate collisions.
- What does that last option look like?

- Suppose we have a hash table with *m* slots.
- Unlike a normal hash table, we'll use *two* hash functions. We'll call them h_1 and h_2 .
- Each hash function outputs a slot number in the set
 { 0, 1, 2, ..., m 1 }.
- We'll assume that these hash functions are truly random, with one constraint:

$h_1(x) \neq h_2(x)$ for any key x.

• The constraint on the hashes being different is pretty easy to achieve both in theory and in



- **The Rule:** Any item xmust either be at position $h_1(x)$ or position $h_2(x)$ in the table.
- Lookups take worstcase O(1) time, since only two locations need to be checked.
- Deletions take *worst-case* O(1) time, since only two locations need to be checked.



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- If that happens, perform a *rehash* by choosing a new h₁ and h₂ and inserting all elements back into the table.
- Multiple rehashes might be necessary before this succeeds – do you see why?



How efficient is cuckoo hashing?
Pro tip: When analyzing a data structure, it never hurts to get some empirical performance data first.



Suppose we have *m* slots and store *n* total elements. What is the probability that all the insertions succeed, as a function of the *load factor* $\alpha = n/m$?



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Suppose we store *n* total elements in a table with *m* slots, where $n < \frac{1}{2}m$.

How many total displacements occur across all insertions?

Goal: Show that insertions take expected time O(1), under the assumption that $n = \alpha m$ for some $\alpha < \frac{1}{2}$.

Analyzing Cuckoo Hashing

- The analysis of cuckoo hashing is more difficult than it might at first seem.
- **Challenge 1:** We may have to consider hash collisions across multiple hash functions.
- **Challenge 2:** We need to reason about chains of displacement, not just how many elements land somewhere.
- To resolve these challenges, we'll need to bring in some new techniques.



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- Each table slot is a node.
- Each element is an edge linking the slots where it can be placed.
- An item's position in the table is denoted with a dot at the end of the line.
- Each node has at most one dot touching it.



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No cycles: The CC containing our edge is a tree. A tree with k nodes has k - 1 edges.

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One cycle: We've added an edge, giving k nodes and k edges.

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Two cycles: There are k nodes and k+1 edges. There are too many edges to place at most one item per node.

- A connected component of a graph is called *complex* if it contains two or more cycles.
- Theorem: Insertion into a cuckoo hash table succeeds if and only if the resulting cuckoo graph has no complex connected components.



How big are the connected components in the cuckoo graph?

(This tells us how much work we do on a successful insertion.)

What is the probability that a connected component in the cuckoo graph is complex?

(This lets us see how much time we should expect to spend rehashing.)

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The Erdős-Rényi model

Random Graph Evolution

- Consider a graph with V nodes and no edges.
- Incrementally add *E* edges to the graph, each chosen uniformly at random, possibly with repetition.
- **Question:** What properties will this graph (probably) have?



Random Graph Evolution

- *Claim:* The phenomena we're observing with cuckoo hashing are, in large part, due to properties of random graphs.
- Good News: This is a wellstudied field! All the results we need were first proved by Erdős and Rényi in 1960.
- This model of incrementally constructing a graph is therefore called the *Erdős-Rényi* model.





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Sizing a Connected Component

- **Goal:** Show that if $E < \alpha V$ for some $\alpha < \frac{1}{2}$, then the expected size of a CC in a randomly-built graph is O(1).
- This seems hard, so let's step away from random graphs for a moment.
- Suppose you have a graph G and a node v in the graph.
- What algorithms might you use to determine the size of the connected component containing v?




































































- **Original Problem:** Figure out the size of the CC containing some initial node *v*.
- **Equivalent Problem:** Figure out the number of nodes in the BFS tree rooted at *v*.
- Framing this as a tree problem turns out to make it easier:
 - We can think about, for each level in the tree, how many nodes are in that level.
 - We can then sum that up across levels to get the number of nodes in the CC.
- "All" we need to do now is apply this to a randomly-built graph.

Sizing a Connected Component

- Pick a starting node for our BFS.
- We want to model how many children it has in the BFS tree.
 - There are *E* total edges.
 - Each edge has a $^{2}/_{V}$ chance of touching our node.
- So this node's number of children is a Binom(E, ²/_V) random variable.



Sizing a Connected Component

- Each new node kinda sorta ish also touches a number of new nodes that can be modeled as a Binom(E, $^{2}/_{V}$) variable.
 - This ignores doublecounting nodes.
 - This ignores existing edges.
 - This ignores correlations between edge counts.
- However, this conservatively bounds the number of new nodes visited in the next BFS step.



- **Idea:** Count nodes in a connected component by simulating a BFS tree, where the number of children of each node is a Binom(E, 2/V) variable.
 - Begin with a root node.
 - Each node has children distributed as a Binom(E, $^{2}/_{V}$) variable.
- *Question:* How many total nodes will this simulated BFS discover before terminating?



- Denote by X_k the number of nodes at level k. This gives a series of random variables X₀, X₁, X₂,
- These variables are defined by the following randomized recurrence relation:

$$X_0 = 1$$
 $X_{k+1} = \sum_{i=1}^{X_k} \xi_{i,k}$

• Here, each $\xi_{i,k}$ is an i.i.d. Binom(E, 2/V) variable.



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- Observation: On expectation, each node has ^{2E}/_V children.
- The "expected branching factor" of the tree is ${}^{2E}/_{V}$, which is less than 1.
- How many nodes are there in the tree, assuming each layer has the expected number of nodes?



There is always one node here.

On expectation, we'd find ${}^{2E}/_V$ nodes here.

On expectation, we'd find $({^{2E}/_V})^2$ nodes here.

On expectation, we'd find $({^{2E}/_V})^3$ nodes here.







$$\mathbf{E}[X_{k+1}] = \mathbf{E}[\sum_{i=1}^{X_k} \boldsymbol{\xi}_{i,k}]$$

This is a sum of a random number of terms, so we can't use linearity of expectation.

However, we can use the *law of total expectation:*

$$\mathbf{E}[X] = \sum_{j} \mathbf{E}[X \mid Y = j] \cdot \mathbf{Pr}[Y = j]$$

$$X_{0} = 1$$

$$X_{k+1} = \sum_{i=1}^{X_{k}} \xi_{i,k}$$

$$\xi_{i,k} \sim \operatorname{Binom}(E, \frac{2}{V})$$

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However, we can use the *law of total expectation:*

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$$= \sum_{j=0}^{\infty} E\left[\sum_{i=1}^{X_{k}} \xi_{i,k} \mid X_{k} = j\right] \cdot \Pr\left[X_{k} = j\right]$$


$$\begin{split} \mathbf{E}\left[X_{k+1}\right] &= \mathbf{E}\left[\sum_{i=1}^{X_k} \xi_{i,k}\right] \\ &= \sum_{j=0}^{\infty} \mathbf{E}\left[\sum_{i=1}^{X_k} \xi_{i,k} \mid X_k = j\right] \cdot \Pr\left[X_k = j\right] \\ \end{split}$$

Well, that
makes things
easier!



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These random variables are independent – one represents the number of nodes in a particular layer. One represents the number of children that a specific node might have.

$$X_{0} = 1$$

$$X_{k+1} = \sum_{i=1}^{X_{k}} \xi_{i,k}$$

$$\xi_{i,k} \sim \operatorname{Binom}(E, \frac{2}{V})$$

$$E[X_{k+1}] = E[\sum_{i=1}^{X_k} \xi_{i,k}]$$

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$$\begin{bmatrix} X_{k+1} \end{bmatrix} = \mathbb{E}\left[\sum_{i=1}^{X_k} \xi_{i,k}\right]$$

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$$= \frac{2E}{V} \cdot \mathbb{E}\left[X_k\right]$$



E

$$E[X_{k+1}] = E[\sum_{i=1}^{X_k} \xi_{i,k}]$$

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$$= \frac{2E}{V} \cdot \sum_{j=0}^{\infty} (j \cdot \Pr[X_k = j])$$

$$= \frac{2E}{V} \cdot E[X_k]$$



The Finishing Touches

- On expectation, there are $({}^{2E}/_V)^k$ nodes in layer k of the BFS tree.
- Summing across all layers, on expectation there are $(1 {^{2E}/_V})^{-1}$ total nodes in the BFS tree.
- Assuming $E = \alpha V$, for a fixed constant $\alpha < \frac{1}{2}$, this is O(1) nodes per CC.
- Therefore, in cuckoo hashing, assuming we set $n = \alpha m$ for some $\alpha < \frac{1}{2}$, each insertion touches a CC with expected size O(1), so each insertion does only expected O(1) displacements.
- This explains why the total number of displacements was such a strongly linear plot!

How big are the connected components in the cuckoo graph?

(This tells us how much work we do on a successful insertion.)

What is the probability that a connected component in the cuckoo graph is complex?

(This lets us see how much time we should expect to spend rehashing.) How big are the connected components in the cuckoo graph? (This tells us how much work we do on a successful insertion.)

What is the probability that a connected component in the cuckoo graph is complex?

(This lets us see how much time we should expect to spend rehashing.)



Consider a random (multi)graph G with V nodes and E edges. What is the the probability that every connected component in G is simple, as a function of the ratio E / V?



Consider a random (multi)graph G with V nodes and E edges. What is the the probability that every connected component in G is simple, as a function of the ratio E / V?

Random Graph Theory

- **Theorem:** Let $E = \alpha V$ for some $\alpha < \frac{1}{2}$. Then the probability that any connected component is complex is $O(\frac{1}{V})$.
- **Corollary:** Using cuckoo hashing with *m* slots and $n = \alpha m$ items, the probability that a series of *n* insertions fails is O(1/n), and the expected number of times a rehash is required before it succeeds is O(1).



Random Graph Theory

- Every proof I've seen of this result boils down to a (messy) counting argument of enumerating possible complex CC shapes and evaluating their probabilities.
- (Possibly?) Open Problem: Find a short, simple proof of the result about complex CCs.



The Overall Analysis

- Cuckoo hashing gives worst-case lookups and deletions.
- Insertions are expected O(1).
 - This assumes you periodically double the size of the table and rehash when things get too full.
- The hidden constants are small, and this is a practical technique for building hash tables.

Cuckoo Hashing:

- *lookup*: O(1)
- *insert*: O(1)*
- *delete*: O(1)

* expected, amortized

Improving Our Space Usage

Improving Space Usage

- A cuckoo hash table with *n* elements requires a table of size *n* / α , with $\alpha < \frac{1}{2}$.
- This means at least 50% of the table slots will be empty.
- The root cause is a fundamental property of random graphs; exceeding this threshold makes failure almost certain.
- **Question:** How can we push past this to improve cuckoo hashing space usage?

Improving Space Usage

- Our cuckoo graph and the associated limitations on cuckoo hashing – result from these two assumptions:
 - Each table slot can hold at most one item.
 - Each item can be placed into one of two positions.
- We need to relax these constraints.

Which of these constraints could we relax, and what might it look like if we did?

Answer at <u>https://pollev.com/cs166spr23</u>



Idea 1: Allow each table slot to store multiple items.

Blocked Cuckoo Hashing

- In **blocked cuckoo hashing**, each slot can hold $b \ge 1$ items.
- When inserting an item, place it in one of the two slots it hashes to if there's free space in either.
- If there's no room left, displace a randomly-chosen other element in the slot.
- Increasing b decreases the likelihood that insertions fail, but increases the cost of lookups and deletions.
- b is often chosen so each slot fits cleanly in a cache line, improving performance.







Suppose we insert $n = \alpha m$ elements into a cuckoo hash table with $m/_b$ slots, each of which can hold *b* elements. What is the probability that all insertions succeed?

Blocked Cuckoo Hashing

- Suppose we have a table with m/b slots, each of which hold b items. Assume $n = m\alpha$ and we use two hash functions.
- The thresholds given below show the maximum value of α a blocked cuckoo hash table can use.
- As you can see, modest increases to *b* dramatically increase the space utilization of the table!

	<i>b</i> = 1	<i>b</i> = 2	<i>b</i> = 3	b = 4	<i>b</i> = 5
Theoretical max α	0.500	0.897	0.959	0.980	0.989

Blocked Cuckoo Hashing

- These bounds are derived from the *k-orientability* threshold for random graphs.
- A *k-orientation* of a graph is a way of placing dots on one endpoint of each edge such that each node has at most *k* dots touching it.
 - In cuckoo hashing, this means that each node stores at most *k* items.
- Deriving these bounds requires a deep dive into random graph theory that's above the CS166 pay grade, but which would be great to learn more about on your own!

	<i>b</i> = 1	<i>b</i> = 2	<i>b</i> = 3	<i>b</i> = 4	<i>b</i> = 5
Theoretical max α	0.500	0.897	0.959	0.980	0.989

Idea 2: Use multiple hash functions.

- In *d*-ary cuckoo hashing, we pick an integer $d \ge 2$ and choose *d* different hash functions.
- Each item can be stored in one up to *d* slots, with choices given by the hash functions.
 - You could do extra work to ensure there are *d* separate locations, or be okay with duplicates if the hashes collide.
- To check if an item is in the table, hash it *d* times and see if it's in any of those slots.
- To insert an item, hash it *d* times and place the item in a free slot. If none exists, evict a randomly-chosen item from a slot, place the new item there, and repeat.





Suppose we insert $n = \alpha m$ elements into a hash table with m slots. What is the probability that all insertions succeed?

- With d = 2, each item in the cuckoo hash table can be in one of two locations.
- We can model this as a graph: nodes are slots and edges are items.
- With d > 2, each item can be in more than two locations.
- How do we model this?



- A *hypergraph* is a generalization of a graph.
- It consists of nodes and hyperedges, where a hyperedge can link any number of nodes.
- A *d-regular* hypergraph is one where each hyperedge links exactly *d* nodes.
- With *d*-ary cuckoo hashing, we get a *d*regular cuckoo hypergraph.



- A 1-orientation of a hypergraph is a way of placing a dot on one endpoint of each hyperedge so that each node has at most one dot on it.
- This corresponds of an assignments of items to slots in a *d*-ary cuckoo hash table.



- Below are the 1-orientability thresholds for *d*-regular hypergraphs.
 - These were worked out over a series of papers in the early 2000s.
- In *d*-ary cuckoo hashing, these give the theoretical maximum α where we can store *n* items in a table with *m* slots, where $n = \alpha m$.
- Notice that adding in even a single extra hash function dramatically increases the space efficiency of the table.

	<i>d</i> = 2	<i>d</i> = 3	d = 4	<i>d</i> = 5	d = 6
Theoretical max α	0.500	0.917	0.976	0.992	0.997
In Practice

- Both blocked cuckoo hashing with b > 1 slots per node and *d*-ary cuckoo hashing with d > 2 significantly improve the space usage of cuckoo hashing.
- In practice, blocked cuckoo hashing tends to be faster than *d*-ary cuckoo hashing.
 - While we tend to forget this, evaluating hash functions is costly.
 - Blocked cuckoo hashing has better locality of reference, since there are (typically) only two cache misses.
- What happens if you combine these approaches together? That's something you'll see on the next problem set.

To Summarize

Summary of Cuckoo Hashing

- Cuckoo hashing is a fast and powerful way to build perfect hash tables.
- We can increase the number of hash functions to increase the load factor, though at a cost to lookup and insert times.
- We can increase the number of items per slot to increase the load factor, though at a cost to lookup and insert times.

Major Ideas for Today

- Randomized data structures using multiple hash functions can often be analyzed from a graph-theoretic perspective.
- Many properties of random graphs and hypergraphs exhibit sharp phase transitions.
- Running experiments is a great way to learn more about randomized data structures.

Next Time

- **TBD!**
 - Fill out the online form to list your preferences for future topics.
 - I'm excited to see what you want to learn!