## Cuckoo Hashing

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- Perfect hashing: Do something clever with multiple hash functions to eliminate collisions.
- What does that last option look like?


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- Maintain two tables, each of which has $m$ elements.
- We choose two hash functions $h_{1}$ and $h_{2}$ from $\mathscr{U}$ to [m].
- Every element $x \in \mathscr{U}$ will either be at position $h_{1}(x)$ in the first table or $h_{2}(x)$ in the second.
- We'll talk about hash strength later; for now, assume truly random hash functions.

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- 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 store $n$ total elements in two tables of $m$ slots each. What's probability all insertions succeed, assuming $m=\alpha n$ ?


Suppose we store $n$ total elements with $m=(1+\varepsilon) n$.
How many total displacements occur across all insertions?

Goal: Show that insertions take expected time $O(1)$, under the assumption that $m=(1+\varepsilon) n$ for some $\varepsilon>0$.

## 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.



## The Cuckoo Graph

- The cuckoo graph is a bipartite multigraph derived from a cuckoo hash table.
- Each table slot is a node.
- Each element is an edge.
- Edges link slots where each element can be.
- Each insertion introduces a new edge into the graph.



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## The Cuckoo Graph

- Claim 1: If $x$ is inserted into a cuckoo hash table, the insertion succeeds if the connected component containing $x$ contains either no cycles or only one cycle.

We either stabilize inside the cycle, avoid the cycle, or get kicked out of the cycle.

## The Cuckoo Graph

- Claim 2: If $x$ is inserted into a cuckoo hash table, the insertion fails if the connected component containing $x$ contains more than one cycle.


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## The Cuckoo Graph

- Claim 2: If $x$ is inserted into a cuckoo hash table, the insertion fails if the connected component containing $x$ contains more than one cycle.


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

## The Cuckoo Graph

- 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? 

## What is the probability that an insert fails?

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 an insert fails?

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 an insert fails?

(This lets us determine how much average work we do on an insertion.)

## Step One: Sizing Connected Components

## Analyzing Connected Components

- The cost of inserting $x$ into a cuckoo hash table is proportional to the size of the CC containing $x$.
- Question: What is the expected size of a CC in the cuckoo graph?

Idea: Count the number of nodes in a connected component by simulating a BFS.

Pick some starting table slot.
There are $n$ elements in the table, so this graph has $n$ edges.
Assume, for now, that our hash functions are truly random.
Each edge has a ${ }^{1 / m}$ chance of touching this table slot.
The number of adjacent nodes, which will be visited in the next step of BFS, is a Binom ( $n, 1 / m$ ) variable.

Idea: Count the number of nodes in a connected component by simulating a BFS.

Each new node kinda sorta ish also touches a number of new nodes on the other side that can be modeled as a $\operatorname{Binom}(n, 1 / m)$ variable.

This ignores double-counting nodes.

This ignores existing edges.
This ignores correlations between edge counts.
However, it conservatively bounds the next BFS step.

## Modeling the BFS

- Idea: Count nodes in a connected component by simulating a BFS tree, where the number of children of each node is a $\operatorname{Binom}(n, 1 / m)$ variable.
- Begin with a root node.
- Each node has children distributed as a Binom ( $n, 1 / m$ ) variable.
- Question: How many total nodes will this simulated BFS discover before terminating?



## Modeling the BFS

- Denote by $X_{k}$ the number of nodes at level $n$. This gives a series of random variables $X_{0}, X_{1}, X_{2}, \ldots$.
- These variables are defined by the following randomized recurrence relation:

$$
X_{0}=1 \quad X_{k+1}=\sum_{i=1}^{X_{k}} \xi_{i, k}
$$

- Here, each $\xi_{i, k}$ is an i.i.d. Binom( $n, 1 / m$ ) variable.


## Modeling the BFS

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$$

- Here, each $\xi_{i, k}$ is an i.i.d. Binom( $n, 1 / m$ ) variable.

There's always exactly one root node in the BFS tree.

$$
X_{o}=1
$$

$$
X_{2}=4
$$

## Modeling the BFS

- Denote by $X_{k}$ the number of nodes at level $n$. This

- Here, each $\xi_{i, k}$ is an i.i.d. Binom ( $n, 1 / m$ ) variable.


## Modeling the BFS

- Observation: On expectation, each node has $n / m$ children.



## Modeling the BFS



## Modeling the BFS

Lemma: $\mathrm{E}\left[X_{k}\right]=(n / m)^{k}$. Proof Idea: Show that<br>$$
\mathrm{E}\left[X_{k+1}\right]=(n / m) \mathrm{E}\left[X_{k}\right]
$$ and apply induction.




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

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:

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

$$
X_{0}=1
$$

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

$$
\xi_{i, k} \sim \operatorname{Binom}\left(n, \frac{1}{m}\right)
$$

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

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

$$
\xi_{i, k} \sim \operatorname{Binom}\left(n, \frac{1}{m}\right)
$$

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

## Well, that makes things



$$
\begin{aligned}
\mathrm{E}\left[X_{k+1}\right] & =\mathrm{E}\left[\sum_{i=1}^{X_{k}} \xi_{i, k}\right] \\
& =\sum_{j=0}^{\infty} \mathrm{E}\left[\sum_{i=1}^{X_{k}} \xi_{i, k} \mid X_{k}=j\right] \cdot \operatorname{Pr}\left[X_{k}=j\right] \\
\begin{array}{l}
\text { that } \\
\text { things }
\end{array} & =\sum_{j=0}^{\infty} \mathrm{E}\left[\sum_{i=1}^{j} \xi_{i, k} \mid X_{k}=j\right] \cdot \operatorname{Pr}\left[X_{k}=j\right]
\end{aligned}
$$

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

This sum ranges over a fixed number of terms, so we can apply linearity of (conditional) expectation.

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



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

This sum ranges over a fixed number of terms, so we can apply linearity of (conditional) expectation.

$$
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& =\sum_{j=0}^{\infty} \mathrm{E}\left[\sum_{i=1}^{j} \xi_{i, k} \mid X_{k}=j\right] \cdot \operatorname{Pr}\left[X_{k}=j\right] \\
& =\sum_{j=0}^{\infty}\left(\sum_{i=1}^{j} \mathrm{E}\left[\xi_{i, k} \mid X_{k}=j\right]\right) \cdot \operatorname{Pr}\left[X_{k}=j\right]
\end{aligned}
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$$
\begin{aligned}
\mathrm{E}\left[X_{k+1}\right] & =\mathrm{E}\left[\sum_{i=1}^{X_{k}} \xi_{i, k}\right] \\
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\end{aligned}
$$

$$
\begin{aligned}
& \begin{array}{c}
\text { These random variables } \\
\text { are independent -one } \\
\text { represents the number of } \\
\text { nodes in a particular layer. } \\
\begin{array}{c}
\text { One represents the } \\
\text { number of children that a } \\
\text { specific node might have. }
\end{array}
\end{array}=\sum_{j=0}^{\infty} \mathrm{E}\left[\sum_{i=1}^{j} \xi_{i, k} \mid X_{k}=j\right] \cdot \operatorname{Pr}\left[X_{k}=j\right] \\
& \hline
\end{aligned}
$$



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

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

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

$$
\begin{aligned}
& X_{0}=1 \\
& X_{k+1}=\sum_{i=1}^{X_{k}} \xi_{i, k}
\end{aligned}
$$



$$
=\sum_{j=0}^{\infty}\left(\sum_{i=1}^{j} \frac{n}{m}\right) \cdot \operatorname{Pr}\left[X_{k}=j\right]
$$

$\xi_{i, k} \sim \operatorname{Binom}\left(n, \frac{1}{m}\right)$

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

$$
=\frac{n}{m} \cdot \sum_{j=0}^{\infty}\left(j \cdot \operatorname{Pr}\left[X_{k}=j\right]\right)
$$

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

$$
=\frac{n}{m} \cdot \sum_{j=0}^{\infty}\left(j \cdot \operatorname{Pr}\left[X_{k}=j\right]\right)
$$

$$
=\frac{n}{m} \cdot \mathrm{E}\left[X_{k}\right]
$$

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

$$
=\sum_{j=0}^{\infty} \mathrm{E}\left[\sum_{i=1}^{X_{k}} \xi_{i, k} \mid X_{k}=j\right] \cdot \operatorname{Pr}\left[X_{k}=j\right]
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=\sum_{j=0}^{\infty}\left(\sum_{i=1}^{j} \frac{n}{m}\right) \cdot \operatorname{Pr}\left[X_{k}=j\right]
$$

$$
=\frac{n}{m} \cdot \sum_{j=0}^{\infty}\left(j \cdot \operatorname{Pr}\left[X_{k}=j\right]\right)
$$

$$
=\frac{n}{m} \cdot \mathrm{E}\left[X_{k}\right]
$$

Lemma 1: $\mathrm{E}\left[X_{k}\right]=(n / m)^{k}$.

## (Induction and conditional expectation.)

Lemma 2: $\mathrm{E}\left[\sum_{i=0}^{\infty} X_{i}\right]=\frac{1}{1-\frac{n}{m}}$.
(Linearity of expectation; sum of a geometric series.)


## The Story So Far

- The expected size of a connected component in the cuckoo graph is $\mathrm{O}(1)$.
- Therefore, each successful insertion takes expected time O(1).
- Question: What happens in an unsuccessful insertion? And what does that do for our expected cost of any insertion?


## Step Two: <br> Exploring the Graph Structure

## Exploring the Graph Structure

- Cuckoo hashing will always succeed in the case where the cuckoo graph has no complex connected components.
- If there are no complex CC's, then we will not get into a loop and insertion time will depend only on the sizes of the CC's.
- It's reasonable to ask, therefore, how likely we are to not have complex components.


## Exploring the Graph Structure

- Question: What is the probability that a randomly-chosen bipartite multigraph with $2 m$ nodes and $n$ edges will contain a complex connected component?
- Directly answering this question is challenging and requires some fairly detailed combinatorics.
- However, there's a clever technique we can use to bound this probability indirectly.


Insertion fails if we have a complex connected component. What specifically happens in that case?


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Insertion fails if we have a complex connected component. What specifically happens in that case?


> We're right back where we started. This pattern will continue indefinitely.

Insertion fails if we have a complex connected component. What specifically happens in that case?


> Question: What's the probability that we end up with a configuration like this one?

Insertion fails if we have a complex connected component. What specifically happens in that case?


This next proof comes from a CS166 final project by Noah Arthurs, Joseph Chang, and Nolan Handali. It's inspired by another argument due to Charles Chen (another Stanford student), which is a modification of one by Sanders and Vöcking, which was an improvement of one by Pagh and Rodler.
Key idea: Use a traditional, CS109-style counting argument. Admittedly, it's a nontrivial counting argument, but it's a counting argument nonetheless!

Insertion fails if we have a complex connected component. What specifically happens in that case?

Ways to split $k$ nodes into $C_{1}, l_{1}, C_{2}$, and $l_{2}$. (upper bound)
Ways to pick $k$ nodes (table slots) given the first is $h_{1}(x)$.

Ways to assign $k$
keys to those slots. (upper bound)
Ways $h_{1}$ and $h_{2}$ can be

$$
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right)
$$

$$
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right)
$$

$$
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right)=\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right)
$$

$$
\begin{aligned}
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right) & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
& =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k-2}\right)
\end{aligned}
$$

$$
\begin{aligned}
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right) & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
& =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k-2}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k}\right)
\end{aligned}
$$

$$
\begin{aligned}
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right) & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
& =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k-2}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}(k+1)^{4}\left(\frac{n}{m}\right)^{k}
\end{aligned}
$$

$$
\begin{aligned}
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right. & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
& =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k-2}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k}\right) \\
m=(1+\varepsilon) n & =\frac{1}{m^{2}} \sum_{k=1}^{n}(k+1)^{4}\left(\frac{n}{m}\right)^{k}
\end{aligned}
$$

$$
\begin{aligned}
\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right) & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
& =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k-2}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{-k}\right) \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n}(k+1)^{4}\left(\frac{n}{m}\right)^{k} \\
& =\frac{1}{m^{2}} \sum_{k=1}^{n} \frac{(k+1)^{4}}{(1+\varepsilon)^{k}}
\end{aligned}
$$

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\sum_{k=1}^{n}\left(\frac{(k+1)^{4} m^{k-1} n^{k}}{m^{2 k} m}\right) & =\sum_{k=1}^{n}\left((k+1)^{4} n^{k} m^{k-1-2 k-1}\right) \\
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Numerator grows polynomially as a function of $k$.

Denominator grows exponentially as a function of $k$.
$=\frac{1}{m^{2}} \sum_{k=1}^{n}(k+1)^{4}\left(\frac{n}{m}\right)^{k}$
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& \begin{array}{c}
\text { Tumerator grows } \\
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Question 1: What is the probability at least one insert fails if we do $n$ total insertions?

The probability that a single insertion fails is $O\left(1 / m^{2}\right)$ if $m=(1+\varepsilon) n$.

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The probability that a single insertion fails is $\mathrm{O}\left(1 / \mathrm{m}^{2}\right)$ if $m=(1+\varepsilon) n$.

If an insertion fails, we rehash by building a brand-new table, with new hash functions, and inserting all old elements.

It's possible that, when we do a rehash, one of the insertions fails. Therefore, we keep rehashing until we find a working table.

## Question 2: On

 expectation, how many rehashes are needed per insertion?
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Let $X$ be a random variable counting the number of rehashes assuming at least one rehash occurs.
$X$ is geometrically distributed with success probability

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E[\#rehashes]
$=\mathrm{E}[X] \cdot \operatorname{Pr}[\#$ rehashes $>0]$

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Question 3: What is the expected cost of an insertion into a cuckoo hash table?


The expected number of rehashes on any insertion is $\mathrm{O}\left(1 / \mathrm{m}^{2}\right)$.

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## $\mathbf{O}(\mathbf{1})+\mathbf{O}\left(\mathbf{1} / \mathrm{m}^{\mathbf{2}}\right) \cdot \mathbf{O}(\mathrm{m})$

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## $\mathbf{O ( 1 )}+\mathbf{O ( 1 / m )}$

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Question 3: What is the expected cost of an insertion into a cuckoo hash table?

## O(1)

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## The Overall Analysis

- Cuckoo hashing gives worstcase lookups and deletions.
- Insertions are expected, amortized $\mathrm{O}(1)$.
- The amortization kicks in because we need to periodically double the sizes of the tables as the number of elements increases.
- The hidden constants are small, and this is a practical technique for building hash tables.


## Cuckoo Hashing:

- lookup: O(1)
- insert: $\mathrm{O}(1)^{*}$
- delete: $\mathrm{O}(1)$
* expected, amortized

More to Explore

## Hash Function Strength

- We analyzed cuckoo hashing assuming our hash functions were truly random. That's too strong of an assumption.
- What we know:
- $O(\log n)$-independence is sufficient for expected $\mathrm{O}(1)$ insertion time, but 6-independence isn't.
- The simplest 2 -independent family of hash functions (polynomial hashing) are terrible for cuckoo hashing.
- Some simple classes of 3-independent hash functions (tabulation hashing) perform well both theoretically and practically.
- Open problem: Determine the strength of hash function needed for cuckoo hashing to work efficiently.


## Multiple Tables

- Cuckoo hashing works well with two tables. So why not $3,4,5, \ldots$, or $k$ tables?
- In practice, cuckoo hashing with $k>2$ tables leads to better memory efficiency than $k=2$ tables:
- The load factor can increase substantially; with $k=3$, it's only around $\alpha=0.91$ that you run into trouble with the cuckoo graph.
- Displacements are less likely to chain together; they only occur when all hash locations are filled in.
- Open problem: Determine where these phase transition thresholds are for arbitrary $k$.


## Increasing Bucket Sizes

- What if each slot in a cuckoo hash table can store multiple elements?
- When displacing an element, choose a random one to move and move it.
- This turns out to work remarkably well in practice, since it makes it really unlikely that you'll have long chains of displacements.
- Open problem: Quantify the effect of larger bucket sizes on the overall runtime of cuckoo hashing.


## Restricting Moves

- Insertions in cuckoo hashing only run into trouble when you encounter long chains of displacements during insertions.
- Idea: Cap the number of displacements at some fixed factor, then store overflowing elements in a secondary hash table.
- In practice, this works remarkably well, since the auxiliary table doesn't tend to get very large.
- Open problem: Quantify the effects of "hashing with a stash" for arbitrary stash sizes and displacement limits.


## Other Dynamic Schemes

- There is another famous dynamic perfect hashing scheme called dynamic FKS hashing.
- It works by using closed addressing and resolving collisions at the top level with a secondary (static) perfect hash table.
- In practice, it's not as fast as these other approaches. However, it only requires 2-independent hash functions.
- Check CLRS for details!


## Lower Bounds?

- Open Problem: Is there a hash table that supports amortized O(1) insertions, deletions, and lookups?
- You'd think that we'd know the answer to this question, but, sadly, we don't.


## Next Time

- Beyond Worst-Case Analysis
- Is $\mathrm{O}(\log n)$ the be-all, end-all of BST analysis? (Hint: Betteridge's Law of Headlines)
- Weight-Balanced Trees
- A different way of balancing a tree.
- Finger Search Trees
- Picking up where we left off.
- Iacono's Working Set Structure
- Storing elements in doubly-exponentially-increasing forests.

