# Heterogeneity and Load Balance in Distributed Hash Tables 

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Joint Work with Alex Fabrikant and Ion Stoica
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## The goals

- Distributed Hash Tables partition an ID space among $n$ nodes
- Typically: each node picks one random ID
- Node owns region between its predecessor and its own ID
- Some nodes get $\log n$ times their fair share of ID space
- Goal 1: Fair partitioning of ID space
- If load distributed uniformly in ID space, then fair partitioning $\Rightarrow$ load balanced system
- Goal 2: Fair partitioning when node capacities are heterogeneous
- Goal 3: Use heterogeneity to our advantage to reduce route length in overlay that connects nodes


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## Model \& performance metric

- $n$ nodes
- Each node $v$ has a capacity $c_{v}$ (e.g. bandwidth)
- Average capacity is 1 , total capacity $n$
- Share of node $v$ is

$$
\operatorname{share}(v)=\frac{\text { fraction of ID space that } v \text { owns }}{c_{v} / n}
$$

- Want low maximum share
- Perfect partitioning has max. share $=1$.


## The virtual server solution

- Goal 1: Load balance
- Each node picks $\Theta(\log n)$ IDs (like simulating $\Theta(\log n)$ nodes)
- Maximum share is $O(1)$ with high probability (w.h.p.) in homogeneous system

- Goal 2: Handle heterogeneity
- Node of capacity $c$ simulates $\Theta(c \log n)$ nodes
- Maximum share is $O(1)$ w.h.p. for any capacity distribution



## Problems

- To route between nodes, construct an overlay network
- With $\Theta(\log n)$ IDs, must maintain $\Theta(\log n)$ times as many overlay connections!

- Other proposals use one ID per node, but...
- all require reassignment of IDs in response to churn, and load movement is costly
- none handles heterogeneity directly
- some can't compute node IDs as hash of IP address for security
- some are limited in the achievable quality of load balance
- some are complicated


## Our Approach

- Our solution: Low Cost Virtual Servers
- Pick $\Theta\left(c_{v} \log n\right)$ IDs for node of capacity $c_{v}$ as before...
- ...but cluster them in a random fraction $\Theta\left(\frac{c_{v} \log n}{n}\right)$ of the ID space
- Random starting location $r$
- Pick $\Theta\left(c_{v} \log n\right)$ IDs spaced at intervals of $\approx \frac{1}{n}$ (with random perturbation)
- Ownership of ID space is still in disjoint segments
- Why does this help?


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## Our Approach: Overlay Topology

- When building overlay network, simulate ownership of contiguous fraction $\Theta\left(\frac{c_{v} \log n}{n}\right)$ of ID space

- Routing ends at node simulating ownership of target ID, not real owner
- But clustering of IDs $\Rightarrow$ real owner is nearby in ID space $\Rightarrow$ can complete route in $O(1)$ more hops using successor links


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## Our Approach: Properties

- Works for any ring-based overlay topology; compared to single-ID case,
- Node outdegree increases by at most a constant factor
- Route length increases by at most an additive constant
- Goals 1 \& 2: Load balance \& handling heterogeneity
- Achieves maximum share of $1+\varepsilon$ for any $\varepsilon>0$ and any capacity distribution
- Tradeoff: outdegree depends on $\varepsilon$


## Max Share Proof

Lemma 1 If node $v$ has at least one ID in the ring and $\alpha=\Theta(\log n)$, then ( 1 ) $v$ has between $\alpha c_{v} /\left(\gamma_{c} \gamma_{u}\right)-O(1)$ and $\alpha c_{v} \gamma_{c} \gamma_{u}+O(1)$ IDs w.h.p., and (2) $v$ has at least $\gamma_{d} \alpha(n)-O(1)$ IDs w.h.p.

Proof: (1) Note that, due to the estimaton error parameters, the factor $\gamma_{c}$ lazy updat of $\tilde{c}_{v}$, and the factor 2 lazy update of $n$, we always have $\tilde{c}_{v}$ within a factor $\gamma_{c} \gamma_{u}$ of $c_{v}$ and $\bar{n}$ within a factor $2 \gamma_{n}$ of $n$ w.h.p. Thus, for some constant $k$, the number of IDs that $v$ chooses is at most $\left\lfloor 0.5+\tilde{c}_{v} \alpha(\tilde{n})\right\rfloor \leq \tilde{c}_{v} \alpha(\tilde{n})+O(1) \leq$ $\gamma_{c} \gamma_{u} c_{v} k \log \left(2 \gamma_{n} n\right)+O(1) \leq \gamma_{c} \gamma_{u} \alpha(n)+O(1)$. The lower bound follows larly, if $v$ has decided to stay in the ring, we must have $\tilde{c}_{v} \geq \gamma_{d}$ and the bound follows by the above technique.

We now break the ring into frames of length equal to the smallest spacing parameter $s_{\text {min }}$ used by any node. The following lemma implies that $s_{\min } \geq 1 /\left(2 \gamma_{n} n\right)$ w.h.p.

Lemma 2 Let $\beta=\left(1-\gamma_{c} \gamma_{u} \gamma_{d}\right) /\left(\gamma_{c} \gamma_{u}\right)$. When $\alpha \geq \frac{8 \gamma_{n}}{\beta \varepsilon^{2}} \ln n$, each frame contains at least $(1-\varepsilon) \beta \alpha n s_{\text {min }}-O(1)$ IDs w.h.p. for any $\varepsilon>0$

Proof: Assume that no node has more than one ID in any frame; if this is not the case, we can break the high-capacity nodes for which it is false into multiple "virtual nodes" withou disturbing the rest of the proof.

Consider any particular frame $f$. Let $X_{v}$ be the indicator variable for the event that node $v$ chooses an ID in $f$ and let $X=\sum_{v} X$. We wish to lower-bound $X$. Suppose $v$ chooses $m_{v}$ points. Since $f$ covers a fraction $s_{\min }$ of the ID space, we have $E\left[X_{v}\right]=m_{v} s_{\text {min }}$. By Lemma ??, $m_{v} \geq \alpha c_{v} /\left(\gamma_{c} \gamma_{u}\right)-O(1)$ for nodes $R$ in the ring. Thus,

$$
\begin{aligned}
\mathrm{E}[X] & =\sum_{v \in R} \mathrm{E}\left[X_{v}\right] \\
& \geq \sum_{v \in R} s_{\min }\left(\alpha c_{v} /\left(\gamma_{c} \gamma_{u}\right)-O(1)\right) \quad \text { (Lemma ??) } \\
& \geq-O(1)+\sum_{v \in R} \frac{s_{\min } \alpha c_{v}}{\gamma_{c} \gamma_{u}} \\
& =-O(1)+\frac{s_{\min } \alpha}{\gamma_{c} \gamma_{u}} \sum_{v \in R} c_{v} \\
& \geq-O(1)+\frac{s_{\min \alpha}}{\gamma_{c} \gamma_{u}} \cdot\left(1-\gamma_{c} \gamma_{u} \gamma_{d}\right) n \quad \text { (Claim ??) } \\
& =\beta \alpha n s_{\min }-O(1),
\end{aligned}
$$

with $\beta$ defined as in the lemma statement. (Note that although Claim ?? was stated in the context of Chord, it applies to our partitioning scheme without modification.) A Chernoff bound tells us that
$\operatorname{Pr}[X<(1-\varepsilon) \mathrm{E}[X]]<e^{-\left(\beta \alpha n s_{m i n}-O(1)\right) \varepsilon^{2} / 2}$

$$
\begin{aligned}
& =O\left(e^{-\beta \alpha n s_{\min } \varepsilon^{2} / 2}\right) \\
& <e^{-\beta \alpha \varepsilon^{2} /\left(4 \gamma_{n}\right)} \quad \text { (Lemma ??) } \\
& =O\left(n^{-2}\right)
\end{aligned}
$$

when $\alpha \geq \frac{8 \gamma_{n}}{\beta \varepsilon^{2}} \ln n$. Again by Lemma ??, there are at $\leq 2 \gamma_{d} n$ frames, so the lemma follows from a union bound over then

Proof: (Of Theorem ??) If node $v$ is discarded its share is 0 , so we need only consider nodes in the ring. Such a node $v$ chooses one ID in each of $m \leq \alpha c_{v} \gamma_{c} \gamma_{u}+O$ (1) frames (Lemma ...).
$X_{1}, \ldots, X_{m}$ be the odes choices of the frames in which they place their IDs. Let domness in the $X_{i}$ s is over the of the ID space owned by each of node $v$ 's IDs. The ranindependently and uniformly at random positions of the nodes' IDs, which are chosen has at least one ID. Thus, the random. By Lemma ??, we may assume that each frame boundary, so $X_{i}$ depends only on the locations of the IDs in its frame and most one frame clockwise preceding frame. Thus, the odd-indexed $X_{i}$ sare mutually independent, as are the time. Consider first the odd-indexed $X_{i}$ s.

Break each frame into $d$ buckets of equal size; we'll pick $d$ later. A bucket is occupied when some node other than $v$ chooses an ID inside it, and is empty otherwise. To analyze the
node $v$ 's share of the ID space, we'll count the number of empty buckets counterclockwisenode $v$ 's share of the ID space, we'll count the number of empty buckets counterclockwise-
following $v$ 's chosen IDs. Define an infinite sequence of randon variables $Y$, each of following $v$ 's chosen IDs. Define an infinite sequence of random variables $Y_{j}$, each of
which will be the indicator variable for the event that a particular bucket is occupied. $Y_{1}$ will correspond to the bucket counterclockwise-following $v$ 's first odd-indexed ID. Suppose $Y_{j}$ corresponds to the $k$ th bucket following $v$ 's $\ell$ th ID. Then we have two cases. (1) If $Y_{j}=0$ $Y_{j+1}$ corresponds to the next bucket following the same ID. (2) Otherwise, $Y_{j+1}$ corresponds to the first bucket following the next odd-indexed ID, i.e. the $(\ell+2)$ th one. If $m / 2<\ell+2$ then we simply set $Y_{j+1}=1$. Thus, the number of zeros in the sequence of $Y_{j}$ 's is the number of buckets entirely owned by $v$ 's $m / 2$ odd-indexed IDs.

With the goal of upper-bounding the number of zeros, we first deal with depen-
and $(1-\varepsilon) \beta s_{\text {min }} n \alpha(n)-O(1)$ IDs for sufficiently large $\alpha$. View $Y_{1}, Y_{2}, \ldots$ as a process. If $Y_{j-1}=1$, then we are in Case (2) and $Y_{j}$ corresponds to a frame independent of those of $Y_{1}, \ldots, Y_{j-1}$, so there are at least $r$ IDs distributed u.a.r. in the frame which may occupy $Y_{j}$ 's bucket. If we are in Case (1) then $Y_{j}$ 's bucket is in the same frame as that of $Y_{j-1}$, which implies that some of the buckets in that frame are empty, in which case there are at least $r$ IDs distributed u.a.r. in a subset of the frame including $Y_{j}$ 's bucket. This
discussion implies that, regardless of the history of the $Y_{j}$ s, the probability that $Y_{j}=1$ is at least $1-(1-1 / d)^{r}$. Formally, we define another sequence of variables $Z_{j}$ which are independent Poisson trials with success probability $p$ to be picked below. For any indeces $j_{1}, \ldots, j_{k}$, we have
$\operatorname{Pr}\left[Y_{j_{1}}=\cdots=Y_{j_{k}}=1\right] \quad=\prod_{\ell=1}^{k} \operatorname{Pr}\left[Y_{j_{\ell}}=1 \mid Y_{j_{1}}=\cdots=Y_{j_{\ell-1}}=1\right.$
$\geq \prod_{\ell=1}^{k}\left(1-\left(1-\frac{1}{d}\right)^{r}\right)$
$\geq\left(1-e^{-r / d}\right)^{k}$
$=\operatorname{Pr}\left[Z_{j_{1}}=\cdots=Z_{j_{k}}=1\right.$
where we have chosen the success probability for the $Z_{j}$ s to be $p=1-e^{-r / d}$. This implies that an upper bound the number of 0 's in the independent $Z_{j}$ sequence is also an apper bound the number of 0 's in the dependent $Y_{k}$ sequence, a fact which we use next.
If we see $m / 2$ ones in the first $x Y_{j}$ s, then by the definition of the sequence, have seen all the zeros, of which there are at most $x-m / 2$. Thus node $v$ will own at most $x-m / 2$ complete buckets, plus 2. $m / 2$ partial buckets (one at each end of the $m / 2$ contiguous sequences of complete buckets), for a total of at most $x+m / 2$ succeses w.h.p. when $x=\frac{m}{2(1)}$. Let $P$ be the number of 1's in the first $x Y$ s,
and let $P^{\prime}$ be the corresponding value for the $Z_{j}$ s. By the above discussion we have $\operatorname{Pr}[P<m / 2] \leq \operatorname{Pr}\left[P^{\prime}<m / 2\right]$, and $\mathrm{E}\left[P^{\prime}\right]=x p=\frac{m}{2(1-\delta)}$, so
$\operatorname{Pr}[P<m / 2] \leq \operatorname{Pr}\left[P^{\prime}<m / 2\right]$
$=\operatorname{Pr}\left[P^{\prime}<(1-\delta) \cdot \frac{m}{2(1-\delta)}\right]$
$\leq e^{-\frac{m \delta^{2}}{4(1-\delta)}}$ (Chernoff bound)
$\leq O\left(e^{-\frac{\gamma_{d} \alpha \delta^{2}}{4(1-\delta)}}\right)$ (Lemma ?? part (2)
$=O\left(n^{-2}\right)$
when $\alpha \geq \frac{8(1-\delta)}{\gamma_{d} \delta^{2}} \ln n$. In this case, counting now both odd- and even-indexed points, node $v$ owns at most $m+\frac{m}{p(1-\delta)}$ buckets, each of size $s_{\min } / d$. Normalizing by $v$ 's fair share $c_{v} / n$, we have

$$
\operatorname{share}(v) \leq \frac{1}{c_{v} / n} \cdot\left(\frac{m s_{\min }}{d}+\frac{m s_{\min }}{d p(1-\delta)}\right)
$$

Recall that $d$ is arbitrary. Taking the limit as $d \rightarrow \infty$, we have $d p \rightarrow r=$ $1-\varepsilon) \beta s_{\min } n \alpha(n)-O(1)$ so

$$
\begin{aligned}
\operatorname{share}(v) & \leq \frac{1}{c_{v} / n} \cdot \frac{m s_{\min }}{(1-\delta)\left((1-\varepsilon) \beta s_{\min } n \alpha(n)-O(1)\right)} \\
& \leq \frac{1}{c_{v}} \cdot \frac{m}{(1-\delta)(1-\varepsilon)\left(1-\varepsilon^{\prime}\right) \beta \alpha(n)} \\
& \leq \frac{1}{c_{v}} \cdot \frac{\alpha(n) c_{v} \gamma_{c} \gamma_{u}+O(1)}{(1-\delta)(1-\varepsilon)\left(1-\varepsilon^{\prime}\right) \beta \alpha(n)} \quad \text { (Lemma ?? part (1)) } \\
& \leq \frac{\left(1+\varepsilon^{\prime \prime}\right)\left(\gamma_{c} \gamma_{u}\right)^{2}}{(1-\delta)(1-\varepsilon)\left(1-\gamma_{c} \gamma_{u} \gamma_{d}\right)}
\end{aligned}
$$

with probability $1-O\left(n^{-2}\right)$ for any $\varepsilon^{\prime}, \varepsilon^{\prime \prime}>0$ and sufficiently large $n$, so by a
union bound, this is true of all nodes w.h.p. Finally, we require that $\alpha$ is the maximum of the
requirement given above and that of Lemma ??; setting $\delta=\varepsilon$ for convenience of presen-
tation, we have $\max \left\{\frac{8(1-\varepsilon) \ln n}{\gamma_{d} \varepsilon^{2}}, \frac{8 \gamma_{n} \gamma_{c} \gamma_{u} \ln n}{\left(1-\gamma_{c} \gamma_{u} \gamma_{d}\right) \varepsilon^{2}}\right\} \leq \frac{8 \gamma_{n} \gamma_{c} \gamma_{u} \ln n}{\left(1-\gamma_{c} \gamma_{u} \gamma_{d}\right) \gamma_{d} \varepsilon^{2}}$, as
required by the theorem.

## Simulation: Maximum share



- Parameter: $\alpha=$ number of virtual servers per unit capacity
- Homogeneous capacities shown here
- Chord with $\alpha=1$ increases to maximum share $\approx 13.7$.


## Simulation: Degree



Degree of a node $=$ number of links to other nodes

## Exploit Heterogeneity (Goal 3)

- Even high-capacity nodes have a single set of overlay links
- Make use of unused capacity: pick denser set of links
- In Chord with $\alpha=1: ~ \Theta\left(c_{v} \log n\right)$ total outlinks
- $\Theta(\log n)$ links in $\Theta\left(c_{v}\right)$ finger tables (one per virtual server)
- In our scheme: $\Theta\left(c_{v} \log n\right)$ total outlinks
- ... all in one dense finger table
- More structured topology $\Rightarrow$ reduced route length


## Simulation: Effect of heterogeneity



- SGG capacity distribution from real Gnutella hosts
- Asymptotic route lengths compared to homogeneous case Chord: $\leq 23 \%$ shorter $\quad Y_{0}: \geq 55 \%$ shorter


## Conclusion

- Advantages
- Simple way to achieve good load balance at low cost
- Compatible with any ring-based overlay
- Adds flexibility in neighbor selection to any overlay
- Takes advantage of heterogeneity to reduce route length
- Disadvantages
- Some additional overhead, especially when particularly good balance desired
- Will incur additional load movement when number of nodes or average capacity changes by a constant factor
- Question: where else does heterogeneity help distributed systems?

Backup slides

## Simulation: Max Share vs. Capacity Distribution



## Simulation: Effect of heterogeneity



Route length vs. capacity distribution in a 16,384 -node system.

