

# Distributed Storage Networks and Computer Forensics 8 Analysis of DHT

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## **Distributed Hash-Table (DHT)**

0

23

f(23)=1

0

### **Pure (Poor) Hashing**

f(1)=4

5

6

4

2

5

#### Hash table

- does not work efficiently for inserting and deleting
- Distributed Hash-Table
  - servers are "hashed" to a position in an continuos set (e.g. line)
  - data is also "hashed" to this set
- Mapping of data to servers
  - servers are given their own areas depending on the position of the direct neighbors
  - all data in this area is mapped to the corresponding server
- Literature
  - "Consistent Hashing and Random Trees: Distributed Caching Protocols for Relieving Hot Spots on the World Wide Web", David Karger, Eric Lehman, Tom Leighton, Mathhew Levine, Daniel Lewin, Rina Panigrahy, STOC 1997





## Entering and Leaving a DHT

- Distributed Hash Table
  - devices are hashed to to position
  - blocks are hashed according to the ID
- When a device is added
  - only blocks from neighbors have to be moved
- When a device is deleted
  - blocks are moved only to the neighbors



### **Holes and Dense Areas**



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### Size of Holes

#### • Theorem

 If n elements are randomly inserted into an array [0,1[ then with constant probability there is a "hole" of size Ω(log n/n), i.e. an interval without elements.

#### Proof

- Consider an interval of size log n / (4n)
- The chance not to hit such an interval is (1-log n/(4n))
- The chance that n elements do not hit this interval is

$$\left(1 - \frac{\log n}{4n}\right)^n = \left(1 - \frac{\log n}{4n}\right)^{\frac{4n}{\log n}\frac{\log n}{4}} \ge \left(\frac{1}{4}\right)^{\frac{1}{4}\log n} = \frac{1}{\sqrt{n}}$$

- The expected number of such intervals is more than 1.
- Hence the probability for such an interval is at least constant.

 $\frac{1}{4} \frac{1}{4} \log n = 2 \left( \frac{1}{4} \log n \right) \log \frac{1}{4} = 2 \log \left( \frac{1}{4} \log n \right) \log \frac{1}{4}$ (- =) · log m  $= m^{2}$   $= m^$ m

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### **Dense Spots**

#### Theorem

 If n elements are randomly inserted into an array
 [0,1[ then with constant probability there is a dense interval of length 1/n with at least Ω(log n/ (log log n)) elements.

#### Proof

- The probability to place exactly i elements in to such an interval is  $\left(\frac{1}{n}\right)^i \left(1 \frac{1}{n}\right)^{n-i} \binom{n}{i}$
- for i = c log n / (log log n) this probability is at least 1/n<sup>k</sup>
   for an appropriately chosen c and k<1</li>
- Then the expected number of intervals is at least 1

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 $\frac{1}{4} \leq \left(1 - \frac{1}{m}\right)^{m} \leq \frac{1}{k}$   $\frac{1}{4} \leq \left(1 - \frac{1}{m}\right)^{m-1} = \left(1 - \frac{1}{m}\right)^{m}$   $\left(1 - \frac{1}{m}\right)^{m-1} = \left(1 - \frac{1}{m}\right)^{m}$   $\geq \left(\frac{1}{4}\right)^{n-\frac{1}{m}}$ 

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$$\begin{pmatrix} m \\ i \end{pmatrix} = \frac{m!}{i!(n-i)!} = \frac{m \cdot (n-i) \cdot (n-2) - (n-i-n)}{i!}$$

$$\geq \frac{m}{n} \cdot \frac{m-i}{n} \cdot \frac{m-2}{n} - \frac{m-i+1}{n} + \frac{1}{n} \leq \frac{1}{2}$$

$$\geq (1 - \frac{i-n}{m})^{n-i} \cdot \frac{n!}{i!}$$

$$(1 - \frac{i-n}{m})^{n-i} \cdot \frac{n!}{i!}$$

$$\leq (\frac{1}{2})^{(1-\frac{i}{m})} (i-n) \geq (\frac{1}{2})^{(1-\frac{i}{m})} = (\frac{1}{2})^{\frac{1}{2}} = (\frac{1}{2})^{\frac{1}{2}}$$

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< c.logn loglogn loj-n c(1+ln(+ln2))

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### **Averaging Effect**

#### Theorem

 If Θ(n log n) elements are randomly inserted into an array [0,1[ then with high probability in every interval of length 1/n there are Θ(log n) elements.

### Excursion

#### Markov-Inequality

- For random variable X>0 with E[X] > 0:  $\mathbf{P}[X \ge k \cdot \mathbf{E}[X]] \le \frac{1}{k}$
- Chebyshev

$$\mathbf{P}[|X - \mathbf{E}[X]| \ge k] \le \frac{\mathbf{V}[X]}{k^2}$$

- for Variance  $\mathbf{V}[X] = \mathbf{E}[X^2] \mathbf{E}[X]^2$
- Stronger bound: Chernoff

### **Chernoff-Bound**

#### Theorem Chernoff Bound

• Let x<sub>1</sub>,...,x<sub>n</sub> independent Bernoulli experiments with

- 
$$P[x_i = 1] = p$$

- 
$$P[x_i = 0] = 1-p$$

• Let 
$$S_n = \sum_{i=1}^n x_i$$

• Then for all c>0

$$\mathbf{P}[S_n \ge (1+c) \cdot \mathbf{E}[S_n]] \le e^{-\frac{1}{3}\min\{c,c^2\}pn}$$

• For 0≤c≤1

$$\mathbf{P}[S_n \le (1-c) \cdot \mathbf{E}[S_n]] \le e^{-\frac{1}{2}c^2pn}$$

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### **Balls and Bins**



Lemma

#### If m= k n ln n Balls are randomly placed in n bins:

- Then for all c>k the probability that more than c ln n balls are in a bin is at most O(n<sup>-c</sup>) for a constant c'>0.
- 2. Then for all c<k the probability that less than c ln n balls are in a bin is at most  $O(n^{-c'})$  for a constant c'>0.

Proof:

Consider a bin and the Bernoulli experiment B(k n ln n,1/n) and expectation:  $\mu = m/n = k \ln n$ 

1. Case: c>2k 
$$P[X \ge c \ln n] = P[X \ge (1 + (c/k - 1))k \ln n] \le e^{-\frac{1}{3}(c/k - 1)k \ln n} \le n^{-\frac{1}{3}(c-k)}$$

2. Case: k<c<2k  $P[X \ge c \ln n] = P[X \ge (1+(c/k-1))k \ln n]$  $\le e^{-\frac{1}{3}(c/k-1)^2k \ln n} \le n^{-\frac{1}{3}(c-k)^2}$ 

3. Case: cP[X \le c \ln n] = P[X \le (1 - (1 - c/k))k \ln n]
$$\le e^{-\frac{1}{2}(1 - c/k)^2k \ln n} \le n^{-\frac{1}{2}(k - c)^2/k}$$

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### **Concept of High Probability**

#### Lemma

If A(i) holds with **high** probability, i.e. 1-n<sup>-c</sup>, then (A(1) and A(2) and ... and A(n)) with **high** probability, i.e. 1-n<sup>-(c-1)</sup>

#### **Proof:**

- ► For all i:  $P[\neg A(i)] \le n^{-c}$
- Hence:  $P[\neg A(1) \text{ or } \neg A(2) \text{ or } \dots \neg A(n)] \le n \cdot n^{-c}$

 $P[\neg(\neg A(1) \text{ or } \neg A(2) \text{ or } ... \neg A(n))] \le 1 - n \cdot n^{-c}$ 

DeMorgan:

$$P[A(1) \text{ and } A(2) \text{ and } \dots A(n)] \leq 1 - n \cdot n^{-c}$$

### **Principle of Multiple Choice**

- Before inserted check c log n positions
- For position p(j) check the distance a(j) between potential left and right neighbor
- Insert element at position p(j) in the middle between left and right neighbor, where a(j) was the maximum choice
- Lemma
  - After inserting n elements with high probability only intervals of size 1/(2n), 1/n und 2/n occur.

### **Proof of Lemma**

# 1. Part: With high probability there is no interval of size larger than 2/n

follows from this Lemma

#### Lemma\*

Let c/n be the largest interval. After inserting 2n/c peers all intervals are smaller than c/(2n) with high probability

# From applying this lemma for c=n/2,n/4, ...,4 the first lemma follows.

### Proof

#### > 2nd part: No intervals smaller than 1/(2n) occur

- The overall length of intervals of size 1/(2n) before inserting is at most 1/2
- Such an area is hit with probability at most 1/2
- The probability to hit this area more than c log n times is at least

$$2^{-c\log n} = n^{-c}$$

 Then for c>1 such an interval will not further be divided with probability into an interval of size 1/(4m).



# Algorithms and Methods for Distributed Storage Networks 9 Analysis of DHT

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