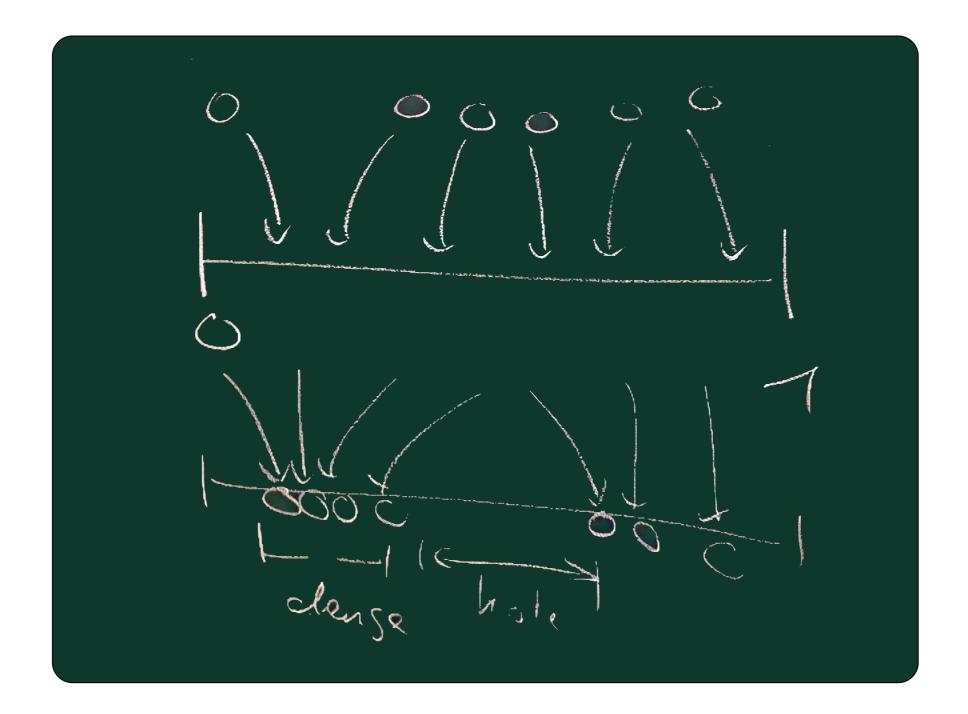


Peer-to-Peer Networks 6. Analysis of DHT

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Holes and Dense Areas





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.

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Proof of Dense Areas

 $\left(\frac{1}{4}\right) \frac{1}{4} \log n = 2 \left(\frac{1}{4} \log n\right) \log \frac{1}{4}$ $= 2^{\left(-\frac{4}{2}\right) \cdot \log n}$ Expectation: $\frac{4m}{\log n} = \frac{1}{\sqrt{m}} + \frac{4\sqrt{m}}{\sqrt{\log m}}$

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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^k
 for an appropriately chosen c and k<1
 - Then the expected number of intervals is at least 1



Proof of Dense Areas

 $i = \frac{c \cdot logn}{log logn}$ ٢ 5~ 1 $Pf: i Balls from m Balls = (1) 1 - 1 n^{-i} (n) = 1$ $fall into an interval of = (1) 1 - 1 n^{-i} (n) = 1$ Size 1 $<math>right = n^{-i} (n) = n^{-i} (n) = 1$

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Proof of Dense Areas

 $\frac{1}{4} \leq \left(1 - \frac{1}{m}\right)^{m} \leq \frac{1}{2}$ $\frac{1}{1 - \frac{1}{m}} = \left(1 - \frac{1}{m}\right)^{m} = \frac{1}{2}$ 1- 1-i-1- 1-i-4 \geq T

Proof of Dense Areas CoNe Freiburg

m. (in. 1). (m-2) ... (m-1-1) γ (n-;)! 1 m-2 n. n-1 n - 1 +n 1 2 h $\frac{1}{m} (i-n) > \left(\frac{1}{c}\right) \left(1-\frac{1}{m}\right) (i-n) \ge \left(\frac{1}{c}\right) \frac{1}{z} = \left(\frac{1}{c}\right) = \left(\frac{1}{c}\right)$

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Proof of Dense Areas

 $\frac{1}{1} = 2$ $\begin{array}{l} \sum_{i+1}^{n} \left(ln \right) & \leq \frac{c \cdot logn}{logb} \left(1 + ln c + ln logn \right) \\ i\left(1 + ln \right) & = \frac{logb}{logb} \left(1 + ln c + ln loglogn \right) \end{array}$ c.logn loglogn 4 len 2 1+h lor loj-n c(1+lmc+lm2)losm

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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.



- 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_1, \ldots, x_n independent Bernoulli experiments with
 - P[x_i = 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

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

• We show
$$\begin{split} \mathbf{P}[S_n \geq (1+c)\mathbf{E}[S_n]] \leq e^{-\frac{\min\{c,c^2\}}{3}pn} \\ & \text{Für t>0:} \\ \mathbf{P}[S_n \geq (1+c)pn] = \mathbf{P}[e^{tS_n} \geq e^{t(1+c)pn}] \\ & k = e^{t(1+c)pn}/E[e^{t\cdot S_n}] \\ & \text{Markov yields:} \\ & \mathbf{P}\left[e^{tS_n} \geq k\mathbf{E}\left[e^{tS_n}\right]\right] \leq \frac{1}{k} \end{split}$$

To do: Choose t appropriately

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• We show
$$\frac{1}{k} \leq e^{-\frac{\min\{c,c^2\}}{3}pn}$$
• where $k = e^{t(1+c)pn}/E[e^{t\cdot S_n}]$

$$= \mathbf{E}\left[\prod_{i=1}^n e^{tx_i}\right]$$
Independence of random variables \mathbf{x}_i

$$= \prod_{i=1}^n \mathbf{E}\left[e^{tx_i}\right]$$

$$= \prod_{i=1}^n (e^0(1-p) + e^tp)$$

$$= (1-p+e^tp)^n$$

$$= (1+(e^t-1)p)^n$$

Show:

$$e^{-t(1+c)pn} \cdot (1+p(e^t-1))^n \le e^{-\frac{\min\{c,c^2\}}{3}pn}$$

where: $t = \ln(1+c) > 0$

$$e^{-t(1+c)pn} \cdot (1+p(e^{t}-1))^{n} \leq e^{-t(1+c)pn} \cdot e^{pn(e^{t}-1)}$$

= $e^{-t(1+c)pn+pn(e^{t}-1)}$
= $e^{-(1+c)\ln(1+c)pn+cpn}$
= $e^{(c-(1+c)\ln(1+c))pn}$

Next to show

$$(1+c)\ln(1+c) \ge c + \frac{1}{3}\min\{c, c^2\}$$



Proof of 1st Chernoff Bound

To show for c>1:

 $(1+c)\ln(1+c) \ge c + \frac{1}{3}c$

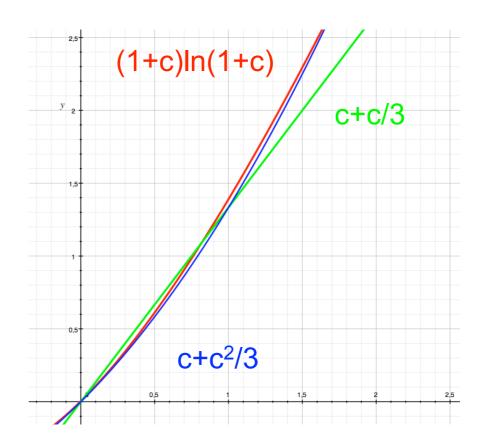
For c=1: $2 \ln(2) > 4/3$

Derivative:

- left side: ln(1+c)
- right side: 4/3
- For c>1 the left side is larger than the right side since

In(1+c)>In (2) > 4/3

 Hence the inequality is true for c>0.



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To show for c < 1:

$$(1+c)\ln(1+c) \ge c + \frac{1}{3}c^2$$

For x>0: $\frac{d\ln(1+x)}{dx} = \frac{1}{1+x} = 1 - x + x^2 - x^3 + x^4 - \dots$

Hence

$$\ln(1+x) = x - \frac{1}{2}x^2 + \frac{1}{3}x^3 - \frac{1}{4}x^4 + \dots$$

By multiplication

$$(1+x)\ln(1+x) = x + \left(1 - \frac{1}{2}\right)x^2 - \left(\frac{1}{2} - \frac{1}{3}\right)x^3 + \left(\frac{1}{3} - \frac{1}{4}\right)x^4 - \dots$$

Substitute (1+c) ln(1+c) which gives for $c \in (0,1)$:

$$(1+c)\ln(1+c) \ge c + \frac{1}{2}c^2 - \frac{1}{6}c^3 \ge c + \frac{1}{3}c^2$$



Theorem Chernoff Bound

- Let x_1, \ldots, x_n independent Bernoulli experiments with
 - P[x_i = 1] = p

- Let
$$S_n = \sum_{i=1}^n x_i$$

- Then for all c>0

$$\mathbf{P}[S_n \geq (1+c) \cdot \mathbf{E}[S_n]] \leq e^{-\frac{1}{3}\min\{c,c^2\}pn}$$
 - For 0≤c≤1

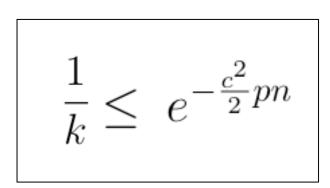
$$\mathbf{P}[S_n \le (1-c) \cdot \mathbf{E}[S_n]] \le e^{-\frac{1}{2}c^2pn}$$



Proof of 2nd Chernoff Bound

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

• For t<0: $P[S_n \le (1-c)pn] = P[e^{tS_n} \ge e^{t(1-c)pn}]$



$$k = e^{t(1-c)pn} / \mathbf{E}[e^{t \cdot S_n}]$$

 \wedge

- Markov yields: $P\left[e^{tS_n} \ge k\mathbf{E}\left[e^{tS_n}\right]\right] \le \frac{1}{k}$
- To do: Choose t appropriately

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Proof of 2nd Chernoff Bound

$$\begin{array}{l|c} \bullet & \mathsf{We show} & \frac{1}{k} \leq e^{-\frac{c^2}{2}pn} \\ \bullet & \mathsf{where} \\ k = e^{t(1-c)pn}/\mathbf{E}[e^{t\cdot S_n}] & = \mathbf{E}\left[e^{t\sum_{i=1}^n x_i}\right] \\ = \mathbf{E}\left[\prod_{i=1}^n e^{tx_i}\right] \\ = \prod_{i=1}^n \mathbf{E}\left[e^{tx_i}\right] \\ = \prod_{i=1}^n (e^0(1-p) + e^tp) \\ = (1-p+e^tp)^n \\ = (1+(e^t-1)p)^n \\ \end{array}$$



$$e^{-t(1-c)pn} \cdot (1+p(e^t-1))^n \le e^{-\frac{c^2}{2}pn}$$

where:

$$t = \ln(1 - c)$$

$$e^{-t(1-c)pn} \cdot (1+p(e^{t}-1))^{n} \leq e^{-t(1-c)pn} \cdot e^{pn(e^{t}-1)} = e^{-t(1-c)pn+pn(e^{t}-1)} \quad 1+x \leq e^{x} = e^{-(1-c)\ln(1-c)pn-cpn}$$

Next to show
$$-c - (1 - c) \ln(1 - c) \le -\frac{1}{2}c^2$$



Proof of 2nd Chernoff Bound

To prove:

$$-c - (1 - c)\ln(1 - c) \le -\frac{1}{2}c^2$$

For c=0 we have equality Derivative of left side: ln(1-c) Derivative of right side: -c

Now

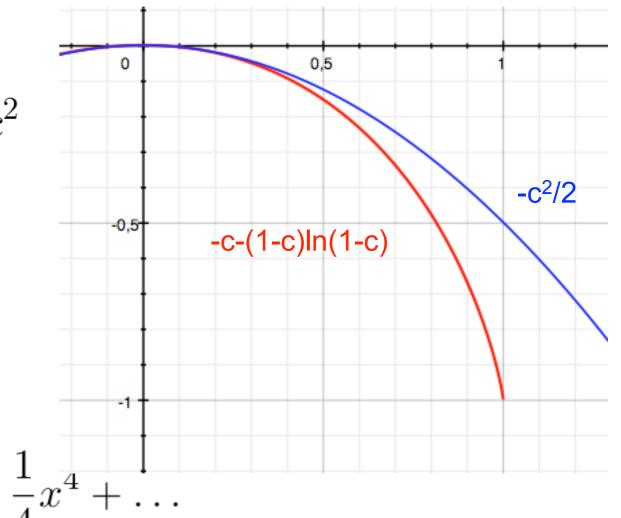
$$\ln(1+x) = x - \frac{1}{2}x^2 + \frac{1}{3}x^3 - \frac{1}{4}x^4 - \frac{1$$

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This implies

$$\ln(1-c) = -c - \frac{1}{2}c^2 - \frac{1}{3}c^3 - \dots < -c$$

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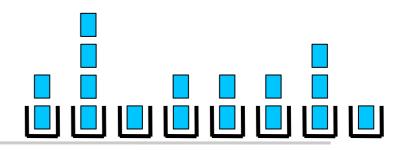
Proof ctd. CoNe Freiburg

 $-c - (1-c)\left(-c - \frac{1}{z}c^{2} - \frac{1}{3}c^{3} - \frac{1}{z}c^{3} - \frac{$ $-c+c+\frac{1}{z}c^{2}+\frac{1}{z}c^{3}+\frac{1}{4}c^{4}+\frac{1}{5}c^{5}$ $-\frac{2}{2}-\frac{1}{3}\frac{3}{3}-\frac{1}{3}\frac{4}{5}-\frac{1}{5}\frac{5}{5}$ $= -\frac{1}{2}c^{2} - (\frac{1}{2} - \frac{1}{3})c^{3} - (\frac{1}{3} - \frac{1}{4})c^{4} - (\frac{1}{4} - \frac{1}{5})c^{5} - (\frac{1}{2} - \frac{1}{3})c^{4} - (\frac{1}{4} - \frac{1}{5})c^{5} - (\frac{1}{2} - \frac{1}{3})c^{5} - ($

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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^{-c'}) for a constant c'>0.
- 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: kP[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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