

# Linear Approximation of Optimal Attempt Rate in Random Access Networks

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**Abstract**—While packet capture has been observed in real implementations of wireless devices randomly accessing shared channels, fair rate control algorithms based on accurate channel models that describe the phenomenon have not been developed. In this paper, using a general physical channel model, we develop the equation for the optimal attempt rate to maximize the aggregate log utility. We use the least squares method to approximate the equation to a linear function of the attempt rate. Our analysis on the approximation error shows that the linear function obtained is close enough to the original with the square of the residuals more than 0.9.

## I. OBTAINING CRITICAL POINT TO MAXIMIZE LOG UTILITIES IN RANDOM ACCESS NETWORKS

We develop an algorithm to obtain optimal rates for log utility fairness [1] using the fixed-point iteration. Let  $S$  be the aggregate log utility. The aggregate utility  $S$  in slotted-Aloha systems is  $\sum_i \log(f_i q_i)$ , where  $f_i$  is the attempt rate of node  $i$ ,  $q_i$  is the success probability of transmissions from node  $i$ . For CSMA/CA systems with a single carrier-sensing ranges,  $S$  is given by  $\sum_i \log\left(\frac{f_i q_i p_{TX}}{(1-p_{TX})T_{SL} + p_{TX}T_{TX}}\right)$ , where  $p_{TX}$  is the probability where any nodes in the network transmit at a time slot, which is simply given by  $p_{TX} = 1 - \prod_{j \in N} (1 - f_j)$ ,  $T_{SL}$  and  $T_{TX}$  are the length of the time slot and transmission time.

The optimal attempt rate allocation makes the differentiation of the aggregate utility equal to zero for all  $i$ . Thus,

$$\begin{aligned} \frac{\partial S}{\partial f_i} &= \frac{1}{f_i} - \frac{1}{1-f_i} \sum_{j \in N} \left( \frac{T_{TX}}{(1-p_{TX})T_{SL} + p_{TX}T_{TX}} - \frac{q_j |i}{q_j} \right) \\ &= \frac{1}{f_i} - \underbrace{\frac{|N|}{G - (1-f_i)}}_A - \underbrace{\frac{|N| - \sum_{j \in N} \frac{q_j |i}{q_j}}{1-f_i}}_B = 0, \end{aligned} \quad (1)$$

where  $G$  is  $\frac{T_{TX}}{(T_{TX} - T_{SL}) \prod_{j \neq i} (1-f_j)}$ . Note that Part A in Equation 1 is omitted if any attempt rates of the nodes equals to 1 or  $T_{TX} = T_{SL}$ .

Especially, for node  $i$ , the optimal attempt rate  $f_i$  satisfies the following:

$$\frac{1-f_i}{f_i} = \sum_j \left( \frac{T_{TX}}{(1-p_{TX})T_{SL} + p_{TX}T_{TX}} - \frac{q_j |i}{q_j} \right). \quad (2)$$

Let  $F^*$  be the optimal attempt rate vector satisfying Equation 1 and function  $g(F^*) = F^*$ . Since function  $g$  is continuous and maps a rate vector to another rate vector,  $g$  has a

fixed point (Brouwer's fixed point theorem [2]). We can further show that  $g$  converges to the fixed point. Thus, if we know  $g$ ,  $F^*$  is obtained by continuously applying  $g$ .

Now, we formulate a function that returns  $f_i$  satisfying Equation 1, given  $f_j$  for  $j \neq i$ . Since  $p_{TX}$  and  $q_j$  need  $f_i$  to compute, Equation 1 is not easy to solve. To get the function to compute  $f_i$ , we first show that Part A in Equation 1 is approximated to a linear function of  $f_i$  as follows:

$$\text{Part A} \approx |N| \left( f_i \mu \left( \prod_{j \neq i} (1-f_j) \right) + \nu \left( \prod_{j \neq i} (1-f_j) \right) \right). \quad (3)$$

We obtain  $\mu$  and  $\nu$  by applying the least squares method. Given  $f_j$  ( $j \neq i$ ), we uniformly sample  $K$  points from the curve  $1/(G - (1-f_i))$ , which is a function of  $f_i$ , and find a linear function that closely approximates the sampled data to minimize the sum of the squares of the residuals between points generated by the function and corresponding sampled points. The computation time of this approximation is  $O(K)$ .

Part B in Equation 1 is approximated by:

$$\begin{aligned} \text{Part B} &= \sum_j \frac{q_j |i - q_j |i}{f_i q_j |i + (1-f_i) q_j |i} \\ &\approx \sum_j \left( \theta(q_j |i, q_j |i) f_i + \phi(q_j |i, q_j |i) \right). \end{aligned} \quad (4)$$

After linear approximation, we have a quadratic formula for  $f_i$  from Equation 1 as follows:

$$\frac{1}{f_i} = a f_i + b, \quad (5)$$

where  $a = |N| \mu \left( \prod_{j \neq i} (1-f_j) \right) + \sum_j \theta(q_j |i, q_j |i)$ ,  $b = |N| \nu \left( \prod_{j \neq i} (1-f_j) \right) + \sum_j \phi(q_j |i, q_j |i)$ .

From the quadratic formula,  $f_i$  is finally given by:

$$f_i = \begin{cases} \frac{-b + \sqrt{b^2 + 4a}}{2a} & (\text{if } a > 0) \text{ and} \\ \min\left(\frac{1}{b}, 1\right) & (\text{if } a = 0), \end{cases}$$

where  $\min(\infty, 1) = 1$ . Assuming  $q_j$  and  $q_j |i$  are known, the total computation time of  $f_i$  is  $O(K)$ .

It is easy to see that  $G$  in Part A of Equation 1 is greater than 1. A linear approximation to Part A with  $f_i$  is prone to larger error as  $G$  gets closer to 1. We, however, show the approximation error is small enough. The value of

$\prod_{j \neq i} (1 - f_j)$  is maximized when all  $f_j$  is the minimum. From Equation 2, the possible minimum of  $f_i$  is obtained when  $q_{j|i}^* = 0$ . Using 802.11 operation parameters to compute  $T_{TX}$  and  $T_{SL}$ ,  $G$  is at least 1.18 for all  $|N| \geq 2$  sending 512-byte packets at 54 Mbps. For all  $|N| \geq 5$ ,  $G$  is more than 1.30. The larger  $|N|$  is, the bigger  $G$  we have. With  $|N| = 2$ , the square of the residuals is around 0.875008831 for  $f_i$  in the range of 0 to 0.5. When  $|N| = 5$ , the square value is beyond 0.927489457. For  $f_i > 0.5$  and  $|N| \geq 2$ , the square value is more than 0.976722922. Note that the square of the residuals is an indicator of how well the linear equation fits. It ranges in value from 0 to 1 and the value 0.927489457 indicates that there is an close correlation in the estimated and actual value of Part A.

To improve the accurate further, we repeat the approximation with different intervals and obtain  $f_i$ . That is, after  $f_i$  is obtained, we find another best-fitting line for a segment of the curve in an interval around the value of  $f_i$ . Then, we again compute a new value of  $f_i$  satisfying Equation 1. Repeating this process several times, we attain the accurate value of  $f_i$ .

In approximation for Part B, the square of the residuals is 0.968194017 for  $f_i$  in the range of 0 to 0.5. Since  $q_{j|\bar{i}}$  and  $q_{j|i}$  are not a function of  $f_i$ , Part B is continuous between  $1/q_{j|i}$  and  $1/q_{j|\bar{i}}$ . It is trivial that the shorter distance between the two end points, the more like a line the graph looks. We can claim that the square of the residuals is maximized when  $q_{j|\bar{i}}$  and  $q_{j|i}$  are farthest away from each other, where  $q_{j|\bar{i}} = 1$  and  $q_{j|i} = 0$  (because  $q_{j|\bar{i}} \geq q_{j|i}$ ). Note that the optimal value of  $f_i$  is typically less than 0.5 in CSMA/CA and slotted-Aloha systems. However, for  $f_i > 0.5$ , we can compute accurate  $\mu$ ,  $\nu$ ,  $\theta$  and  $\phi$  by repeating the approximation with proper intervals.

#### REFERENCES

- [1] F. Kelly, A. Maulloo, and D. Tan, "Rate control in communication networks: shadow prices, proportional fairness and stability," in *Journal of the Operational Research Society*, vol. 49, 1998.
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