# The Impact of Power Amplifier Characteristics on Routing in Random Wireless Networks

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Abstract— Power amplifiers for wireless transmission provide a limited radiated power, and their efficiency depends highly on the operating point. We show that power control and routing strategies in multi-hop wireless networks are strongly affected by these non-ideal amplifier characteristics. For the analysis, we prove that the distances in random networks are governed by a generalized Rayleigh distribution, and we determine the power efficiency of different routing schemes. The main result is that nearest-neighbor routing is highly inefficient if the network has to be connected with high probability.

## I. INTRODUCTION

Energy consumption in multi-hop wireless networks is a crucial issue that needs to be addressed at all the layers of the communication system, from the hardware up to the application. In this paper, we focus on the impact that the characteristics of the power amplifier has on energy-efficient routing strategies. The analysis is based on a Rayleigh fading channel model, and the results demonstrate that the properties of the hardware and the physical channel have a substantial impact on optimum protocol design at the network layer.

Two assumptions are generally made when assessing energy consumption in multi-hop networks: 1) the power consumption is equal (or proportional) to the radiated power; 2) reliable links exist if the receiver is within a certain distance of the transmitter, and interference is taken into account using the same geometric disk abstraction. Such a deterministic "disk model" is used for the analysis of multi-hop packet networks in [1]–[9], thereby ignoring the stochastic nature of the fading channel. Using such models, it is easy to show that, for a path loss exponent of  $\alpha$ , there is an energy gain of  $n^{\alpha-1}$  if a hop over a distance d is split into n hops of distance d/n. However, the volatility of the channel cannot be ignored in wireless networks [10], [11]; the inaccuracy of "disk models" has also been pointed out in [12] and is easily demonstrated experimentally [13].

To overcome some of these limitations of the "disk model", we employ a simple Rayleigh fading link model that relates transmit power, large-scale path loss, and the success of a transmission.

While fading has been considered in the context of packet networks [14], [15], its impact on the network (and higher) layers is largely an open problem. Similarly, non-ideal characteristics of power amplifiers are not usually considered at higher layers. In this paper, we take both fading and amplifier properties into account, and we show that this cross-layer perspective sheds some new light on the routing problem.

# II. THE RAYLEIGH NETWORK MODEL

#### A. The Rayleigh fading link model

We assume a narrowband Rayleigh block fading channel. A transmission from node *i* to node *j* is successful if the SINR  $\gamma_{ij}$  is above a certain threshold  $\Theta$  that is determined by the communication hardware, and the modulation and coding scheme [10]. The SINR  $\gamma$  is a discrete random process given by  $\gamma = \frac{R}{N_0+I}$ . *R* is the received power, which is exponentially distributed with mean  $\overline{R}$ . Over a transmission of distance  $d = ||x_i - x_j||_2$  with an attenuation  $d^{\alpha}$ , we have  $\overline{R} = P_0 d^{-\alpha}$ , where  $P_0$  is proportional to the transmit power<sup>1</sup>.  $N_0$  denotes the noise power, and *I* is the interference power affecting the transmission, *i.e.*, the sum of the received power from all the undesired transmitters.

**Theorem 1** In a Rayleigh fading network, the reception probability  $\mathbb{P}[\gamma \ge \Theta]$  can be factorized into the reception probability of a zero-noise network and the reception probability of a zero-interference network.

*Proof:* Let  $R_0$  denote the received power from the desired source and  $R_i$ , i = 1, ..., k, the received power from k interferers. All the received powers are exponentially distributed, *i.e.*,  $p_{R_i}(r_i) = 1/\bar{R}_i e^{-r_i/\bar{R}_i}$ , where  $\bar{R}_i$  denotes the average received power  $\bar{R}_i = P_i d_i^{-\alpha}$ . The probability of correct reception is (a similar calculation has been carried out in the Appendix of [14] for a network with spreading gain and equal transmit powers for all nodes.)

$$p_r = \mathbb{P}[R_0 \ge \Theta(I + N_0)] = \exp\left(-\frac{\Theta(I + N_0)}{\bar{R}_0}\right) \quad (1)$$

$$= \int_{0}^{\infty} \cdots \int_{0}^{\infty} \exp\left(-\frac{\Theta(\sum_{i=1}^{k} r_{i} + N_{0})}{\bar{R}_{0}}\right)$$
$$\cdot \prod_{i=1}^{k} p_{R_{i}}(r_{i}) \, \mathrm{d}r_{1} \cdots \mathrm{d}r_{k}$$
(2)

$$= \underbrace{\exp\left(-\frac{\Theta N_0}{P_0 d_0^{-\alpha}}\right)}_{p_r^N} \cdot \underbrace{\prod_{i=1}^k \frac{1}{1 + \Theta \frac{P_i}{P_0} \left(\frac{d_0}{d_i}\right)^{\alpha}}}_{p_r^I}.$$
 (3)

<sup>1</sup>This equation does not hold for very small distances. So, a more accurate model would be  $\bar{R} = P'_0 \cdot (d/d_0)^{-\alpha}$ , valid for  $d \ge d_0$ , with  $P'_0$  as the average value at the reference point  $d_0$ , which should be in the far field of the transmit antenna. At 916MHz, for example, the near field may extend up to 3-4ft (several wavelengths).

 $p_r^N$  is the probability that the SNR  $\gamma_N := R_0/N_0$  is above the threshold  $\Theta$ , *i.e.*, the reception probability in a zerointerference network as it depends only on the noise. The second factor  $p_r^I$  is the reception probability in a zero-noise network.

This allows an independent analysis of noise and interference. If the load is light (low interference probability), then SIR $\gg$ SNR, and the noise analysis alone provides accurate results. For high load, a separate interference analysis has to be carried out [16]. Note that *power scaling*, *i.e.*, scaling the transmit powers of all the nodes by the same factor, does not change the SIR ( $p_r^I$  only depends on *power ratios*), but (slightly) increases the SINR. In any case, the noise analysis is relevant, so the main focus of this paper is on the noise.

In a zero-interference network, the reception probability over a link of distance d at a transmit power  $P_0$ , is given by  $p_r := \mathbb{P}[\gamma_N \ge \Theta] = e^{-\frac{\Theta N_0}{P_0 d^{-\alpha}}}$ , therefore

$$P_0 = \frac{d^{\alpha} \Theta N_0}{-\ln p_r}.$$
(4)

Note that for high probabilities, the packet loss probability  $1 - p_r$  is tightly upperbounded by the normalized mean NSR  $\Theta N_0/\bar{R}_0 = \Theta/\bar{\gamma}_N$  [17]. Since  $-\ln p_r \approx 1 - p_r$ , we can also say that the packet loss probability is inversely proportional to the transmit power for high  $p_r$ .

# B. Random networks with uniform distribution

If nodes are distributed uniformly with a density  $\lambda$  in a large network, the probability of finding k nodes in an area A is given by the Poisson distribution [12]

$$\mathbb{P}[k \text{ nodes in } A] = e^{-\lambda A} \frac{(\lambda A)^k}{k!}.$$
(5)

Hence, the positions of the nodes constitute a Poisson point process in the plane<sup>2</sup>. Without loss of generality, we can restrict ourselves to the case  $\lambda = 1$  (unit density), since the product  $\lambda A$  can always be scaled such that  $\lambda = 1$ .

For the routing schemes we consider, we need to determine the distance from one node to its neighboring nodes that lie within a sector  $\phi$ , *i.e.*, within  $\pm \phi/2$  of the source-destination axis (Fig. 1).

**Proposition 1** In a random network with uniform distribution and unit density, the distance R between a node and its nearest neighbor in a sector  $\phi$  is Rayleigh distributed with mean  $\sqrt{\pi/(2\phi)}$ .

*Proof:* Let R be the distance to the nearest neighbor in a sector  $\phi$ . The probability that there is no neighbor in a sector  $\phi$  up to a distance r is the complementary cumulative distribution  $\mathbb{P}[R > r] = e^{-r^2\phi/2}$ , thus the probability density is  $p_R(r) = r\phi e^{-r^2\phi/2}$ , which is a Rayleigh distribution with mean  $\sqrt{\pi/(2\phi)}$  and variance  $2/\phi - \pi/(2\phi) = (4 - \pi)/(2\phi)$ . The distribution of the argument  $\psi$  is uniform between  $-\phi/2$  and  $\phi/2$ .

<sup>2</sup>This can be generalized to higher dimensions if A is the Lebesgue measure of the subset considered.



Fig. 1. Part of a Rayleigh network with the source at the origin and the x-axis pointing towards the destination node. R denotes the distance to the nearest neighbor within a sector  $\phi$  around x, and  $\psi$  is its argument. Hence  $(R, \psi)$  are the polar coordinates of the nearest neighbor within a sector  $\phi$ .

**Definition 1 (Rayleigh network.)** A Rayleigh network *is a large random network with uniformly distributed nodes where the physical channel is subject to Rayleigh fading.* 

**Proposition 2** *The probability density of the distance to the* n*-th nearest neighbor in a sector*  $\phi$  *is* 

$$p_{R_n}(r) = r^{2n-1} \left(\frac{\phi}{2}\right)^n \frac{2}{(n-1)!} e^{-r^2 \phi/2} \tag{6}$$

*Proof:* Let  $S_k$  be the k-th coefficient in the Poisson distribution:  $S_k := (r^2 \phi/2)^k / k!$ . The probability that there are less than n nodes closer than r in the sector  $\phi$  is

$$P_n := \mathbb{P}[0...n-1 \text{ nodes within } r] = \sum_{k=0}^{n-1} S_k e^{-r^2 \phi/2}.$$
 (7)

From  $p_{R_n} = \frac{d}{dr} (1 - P_n)$  we get

$$p_{R_n} = \left(r\phi \sum_{k=0}^{n-1} S_k - \sum_{k=1}^{n-1} \underbrace{\frac{k(r^2\phi/2)^{k-1}}{k!}}_{S_{k-1}} r\phi\right) e^{-r^2\phi/2}.$$
 (8)

The only term that is not cancelled in the two sums is the one at n-1, leading to

$$p_{R_n} = r\phi \cdot \underbrace{S_{n-1} e^{-r^2 \phi/2}}_{\text{Erlang distribution}}, \qquad (9)$$

which is identical to (6).

Since  $p_{R_n}$  is a Rayleigh distribution for n = 1, it can be considered a generalized Rayleigh distribution. Similarly, for a one-dimensional Poisson process, the Erlang distribution is a generalized exponential distribution. So, the transition from one dimension to two dimensions entails a multiplication by  $r\phi$  (that comes from the inner derivative of the exponential part) in the distributions of the node distances. The distributions for  $n = 1, \dots, 8$  are shown in Fig. 2.



Fig. 2. The probability density function of the distances to the *n*-th nearest neighbor for n = 1 (leftmost curve) to n = 8 (rightmost curve) and  $\phi = \pi/2$ . For n = 1, this is a Rayleigh distribution. The mean values are indicated by the dashed lines. For n = 1 and n = 2, the mean values are 1 and 3/2, respectively.

The mean of  $R_n$  is given by

$$\mathbb{E}[R_n] = \frac{\sqrt{2}}{\sqrt{\phi}} \frac{\Gamma(n+1/2)}{\Gamma(n)} = \frac{\sqrt{2}}{\sqrt{\phi}} \frac{(2n)!\sqrt{\pi}}{n!(n-1)!4^n}$$
$$\approx \sqrt{\frac{4(n-1)+\pi}{2\phi}}, \qquad (10)$$

where we have made use of some properties of the  $\Gamma(\cdot)$  function [18] and then derived a highly accurate approximation that shows that the mean distance to the *n*-th neighbor increases with the square root of *n*. The second moment is  $2n/\phi$ , hence the variance is

$$\operatorname{Var}[R_n] = \frac{2n}{\phi} - E[R_n]^2 = \frac{4-\pi}{2\phi}, \qquad (11)$$

which is, interestingly, independent of n.

## **III. POWER AMPLIFIER CHARACTERISTICS**

The most energy efficient operation of an RF power amplifier (PA) is near saturation as this is when the power added efficiency<sup>3</sup> (PAE) is largest. Linear amplification is possible mainly by operating the power amplifier with a small input signal (large backoff) where the energy efficiency of the amplifier is smaller. This characteristic of nonlinear amplifiers makes large power efficiency and bandwidth efficiency hard to achieve.

Depending on the modulation scheme and the specific application, there are situations in which linearity, especially amplitude linearity, can be traded for efficiency and RF power output [19]. Such applications include constant-envelope schemes such as FSK and GMSK, which can tolerate high levels of amplitude distortion, and intermediate cases such as OQPSK and DQPSK systems, which can tolerate significant amounts of amplitude distortion. On the other hand, for modulations with non-constant envelope, amplifier linearity is important. In order to obtain higher PAE, a higher class of amplifiers such as class AB, class B or even class C or switched mode class E or F is often used. However, due to the I/V curves of devices operated in these classes, the amplifier becomes nonlinear. Therefore, high efficiency and high linearity are often contradictory objectives [20], [21].

To avoid discussions of amplifier classes, modulations schemes, different "overdrive" conditions, matching networks and other implementation details, we employ a simple piecewise linear model for the radiated power vs. bias power characteristics of a PA:

$$P_{dc} = \beta P_{TX} + P^* \qquad 0 \leqslant P_{TX} \leqslant P_{\max} \,, \qquad (12)$$

where  $P_{\text{max}}$  is the maximum power that the modulation schemes tolerates (this can be close to saturation or well within the linear region of the PA), and  $P^*$  is the total static power consumption. The slope  $\beta$  normally ranges from 0 to 2. The maximum power efficiency  $\kappa = P_{TX}/P_{dc}$  is reached at  $P_{TX} = P_{\text{max}}$ , where  $\kappa = \frac{1}{\beta + P^*/P_{\text{max}}}$ . The ratio  $P^*/P_{\text{max}}$ ranges from 1/5 to 1.

While such a model can be justified from a hardware point of view, it can also be derived from data sheets of existing PAs and single-chip transceivers manufactured by National Semiconductor, Maxim, RF Micro Devices, Motorola, TriQuist Semiconductor, Agilent, and Mitsubishi designed for 3G, WLAN, and Bluetooth applications [22].

To further simplify the discussion, we define the *generic amplifier*:

**Definition 2 (Generic amplifier.)** A generic amplifier is an amplifier with  $\beta = 1$  and  $P_{\text{max}} = P^*$  in (12). Its maximum efficiency of  $\kappa = 50\%$  is reached at  $P_{TX} = P_{\text{max}}$ .

Note that the  $P_{dc}(P_{TX})$  characteristics of a large number of commercial amplifiers are upperbounded by this generic amplifier curve, *i.e.*, most PAs have a smaller efficiency. With this model, we can state the following result on multi-hop vs. single-hop communication:

**Proposition 3** For the generic amplifier, there is no energy benefit in using multiple hops, if the destination can be reached at  $P_{TX} = P_{\text{max}}$  directly with the desired reliability.

*Proof:* The power consumption for the one-hop case is<sup>4</sup>  $E_1 = 2P_{\text{max}}$ . For the *n*-hop case, we have  $E_n = n(P_{\text{max}} + n^{-\alpha}P_{\text{max}})$ , thus the ratio is

$$\frac{E_n}{E_1} = \frac{n}{2} (1 + n^{-\alpha}), \qquad (13)$$

which is bigger than 1 for any  $\alpha$  for  $n \ge 2$ .

#### **IV. ROUTING SCHEMES**

As shown in Fig. 1, we want to ensure that the source node is connected to a neighbor within a sector  $\phi$  with (high) probability  $p_c$ .

 ${}^{4}$ It is assumed that *E* expresses the energy required to send one packet at a power level *P*.

<sup>&</sup>lt;sup>3</sup>There exist different definitions of PAE. We are considering the one that takes the drive power into account, which results in  $PAE=P_{TX}/P_{dc}$ .



Fig. 3. Contour plot of  $P_{\max}$  as a function of  $p_r$  and  $p_c$  for  $\alpha = 2$  (light) and  $\alpha = 4$  (dark). The spacing between contour lines is 1dB, with increasing  $P_{\max}$  for higher  $p_r$  and/or  $p_c$ . The contours for the different path loss exponents have different absolute values.

**Proposition 4** Let  $p_c$  denote the probability that there is a neighbor with a certain distance d in the sector  $\phi$ . The necessary power to reach this nearest neighbor within a sector  $\phi$  with probability  $p_r$  is

$$P_{\max} = \frac{\Theta N_0}{-\ln p_r} \left(\frac{-2\ln(1-p_c)}{\phi}\right)^{\alpha/2}.$$
 (14)

*Proof:* From (4) follows  $P_{\max} = -N_0 \Theta d_{\max}^{\alpha} / \ln p_r$ , and we get  $d_{\max}^2 = -2 \ln(1-p_c)/\phi$  from  $p_c = \mathbb{P}[R < d_{\max}] = 1 - e^{-d_{\max}^2 \phi/2}$ .

Figure 3 shows a contour plot of  $P_{\max}(p_r, p_c)$ .

If the maximum transmit power is given, the sector  $\phi$  has to satisfy

$$\phi \ge -2\ln(1-p_c) \left(\frac{\Theta N_0}{P_{\max}(-\ln p_r)}\right)^{2/\alpha}.$$
 (15)

The average distance to the nearest neighbor is  $\bar{d} = \sqrt{\pi/(2\phi)}$ . So,  $d_{\text{max}}/\bar{d} = 2\sqrt{-\ln(1-p_c)/\pi}$ , independent of  $\phi$  or  $p_r$ .

# A. Nearest-neighbor routing

Under optimum power control, the mean *backoff* of the generic PA from the optimum operating point is  $(d_{\text{max}}/\bar{d})^{\alpha}$ . For  $\alpha = 2$ , this is about 4 at  $p_c = 95\%$  and 10 at  $p_c = 99.9\%$ .

We define the *average efficiency*  $\bar{\kappa}$  to be the efficiency at a transmit power of  $P = (\bar{d}/d_{\text{max}})^{\alpha} P_{\text{max}}$ . We find

$$\bar{\kappa} = \frac{1}{1 + \left(2\sqrt{\frac{-\ln(1-p_c)}{\pi}}\right)^{\alpha}}.$$
(16)

This relationship is shown in Fig. 4 for  $\alpha = 2, 3, 4$ .



Fig. 4. Average efficiency of the power amplifier as a function of  $p_c$  for  $\alpha=2,3,4.$ 

#### B. Routing to the *n*-th nearest neighbor

We have already determined the distance to the  $n^{\text{th}}$  nearest neighbor in (6). The corresponding cumulative density cannot be analytically solved for the distance, but we notice that all the curves have similar shapes (which is also corroborated by the fact that all distributions have the same variance), and the mean is increasing with  $\sqrt{n}$ . Is it therefore reasonable to assume that, with  $p_c = \mathbb{P}[R_n < d_n]$ , for any given  $p_c$ , q := $d_n - \bar{d}_n$  does not depend on n. Hence, the backoff  $d_n/\bar{d}_n \approx$  $1 + q/\bar{d}_n$  is getting smaller with increasing n. So, it is a good strategy to design the PA such that it can reach over longer distances and transmit to n-th nearest neighbors instead of just nearest neighbors. Due to the additional terms in (7), it is clear that q is slightly decreasing with increasing n. So, if we insert the value for n = 1, we get a lower bound for the efficiency for larger n.

$$\frac{d_n}{d_n} < 1 + \frac{2\sqrt{-\ln(1-p_c)} - \sqrt{\pi}}{\sqrt{4(n-1) + \pi}} \approx 1 + \frac{\sqrt{-\ln(1-p_c)}}{\sqrt{n}}.$$
(17)

The resulting average efficiency is

$$\bar{\kappa}_n > \frac{1}{1 + \left(1 + \sqrt{\frac{-\ln(1-p_c)}{n}}\right)^{\alpha}}.$$
(18)

This bound gets tight with increasing n, and it shows that the efficiency (slowly) approaches 1/2 as n increases.

# C. Optimum routing

Since the efficiency of the PA is optimum at  $P_{\text{max}}$ , the routing scheme should try to identify nodes that are as far away as possible within the radius that allows a reception with probability  $p_r$ . This might be the first, second, third,... or *n*-th neighbor within the sector. In other words, if there are exactly n neighbors within the sector, then the routing algorithm should use the *n*-th nearest one. Hence the probability that the *n*-th nearest neighbor is chosen is given by the Poisson



Fig. 5. Average efficiency for the optimum scheme for  $\alpha = 2$  (top curve)  $\alpha = 3$  and  $\alpha = 4$  (bottom curve).

term  $(d^2\phi/2)^k/k! e^{-d^2\phi/2}$ . We need the density function of the distance to this neighbor:

**Proposition 5** The probability density of the distance to the furthest neighbor within distance  $d_{max}$  in a sector  $\phi$ , given that there is a least one neighbor in the sector, is

$$p_R(r) = \frac{r\phi e^{r^2\phi/2}}{e^{d_{\max}^2\phi/2} - 1}, \quad r \in [0, d_{\max}].$$
(19)

*Proof:* The complementary cumulative distribution  $\mathbb{P}[R > r]$ , conditioned on having at least one node in the sector within distance  $d_{\max}$ , is given by the probability that there is (at least) one node with distance  $r < R \leq d_{\max}$ :

$$\mathbb{P}[R > r] = \frac{1 - e^{-(d_{\max}^2 - r^2)\phi/2}}{1 - e^{-d_{\max}^2\phi/2}}.$$
 (20)

For the mean distance, we get

in Fig. 5.

$$\bar{d} = \mathbb{E}[R] = \frac{d_{\max}e^{d_{\max}^2\phi/2} - c}{e^{d_{\max}^2\phi/2} - 1}$$
(21)

with  $c := \sqrt{\frac{\pi}{2\phi}} \operatorname{erfi}(\frac{d_{\max}}{2}\sqrt{2\phi})$ , where  $\operatorname{erfi}(\cdot)$  is the imaginary error function, *i.e.*,  $\operatorname{erfi}(x) = 2/\sqrt{\pi} \cdot \int_{t=0}^{x} e^{t^2} dt$ .  $\overline{d}$  tends to  $d_{\max}$  with increasing  $d_{\max}$ , hence the efficiency increases with increasing  $p_c$ , in contrast to the other schemes. This is shown

#### V. CONCLUDING REMARKS

Using a Rayleigh fading channel model and a simple power amplifier model that takes into account that the power added efficiency strongly depends on the transmit power, we have shown that the benefits of multi-hop routing vanish completely if the maximum radiated power allows to reach a destination in a single hop. In the case of random networks with uniform distribution, routing schemes that transmit as far as possible clearly outperform nearest-neighbor routing. The optimum strategy is to choose the furthest neighbor (within a certain sector of the axis to the destination) that can be reached with sufficient reliability.

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