

The Effects of Limited Channel Knowledge on Cognitive Radio System Capacity

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Abstract—We examine the impact of limited channel knowledge on the secondary user (SU) in a cognitive radio system. Under a minimum signal-to-interference-and-noise ratio (SINR) constraint for the primary user (PU) receiver, we determine SU capacity under five channel knowledge scenarios. We derive analytical expressions for the capacity cumulative distribution functions and the probability of SU blocking as a function of allowable interference. We show that imperfect knowledge of the PU-PU link gain by the SU-Tx often prohibits SU transmission or necessitates a high interference level at the PU. We also show that errored knowledge of the PU-PU channel is more beneficial than statistical channel knowledge and that imperfect knowledge of the SU-Tx to PU-Rx link has limited impact on SU capacity.

I. INTRODUCTION

A large body of work is now available on various aspects of CR systems, including fundamental information theoretic capacity limits and performance analysis, which often assumes perfect SU-Tx to PU-Rx channel state information (CSI) [1]–[7]. In practice, there is expected to be limited (or no) collaboration between PU and SU systems. Hence, an important question is the impact of the nature of channel knowledge on CR capacity. Several recent contributions have considered imperfect CSI [8]–[14]. In [8], mean and outage capacities along with optimum power allocation policies have been investigated for a CR system in a fading environment with imperfect CSI. Here, probabilistic constraints were employed to maintain an acceptably low probability that interference exceeded some target. In our work, we also use probabilistic constraints, but apply them to a signal-to-interference noise ratio (SINR) target.

This paper differs from the existing literature in several ways. There are four link gains in a two user PU/SU channel to consider and each of them may or may not be perfectly

known at the SU transmitter. Previous studies [8]–[10], [12], [14] have only assumed imperfect knowledge of the SU-Tx to PU-Rx link. Additionally, in previous work, the effect of the interference from the PU-Tx on SU capacity is ignored. Also, we employ the SINR at the PU-Rx to impose probabilistic constraints to protect the PU-Rx, while prior works, with the exception of [10], have considered an interference outage constraint. Finally, we consider several cases where the imperfect CSI manifests itself in the form of statistical channel knowledge (i.e., knowledge of the mean link gains). Such a form of imperfect CSI is attractive from a practical stand point, since obtaining accurate knowledge is almost impossible for some links, such as the PU-Tx to PU-Rx link. Moreover, the mean value does not impose a large system burden as it only requires infrequent updates. Note that the inclusion of PU-Tx to SU-Rx interference and probabilistic constraints enables a rigorous evaluation of the benefits of various types of CSI. In this paper, we establish the following key observations and results:

- In four of the five scenarios considered, we derive analytical expressions for the cumulative distribution function (cdf) of the SU SINR and use it to evaluate the SU capacity cdf.
- For all scenarios, we derive the probability of SU blocking as a function of the permissible interference at the PU-Rx.
- By evaluating our results for a range of system parameters, we demonstrate the importance of accurate knowledge of the PU-Tx to PU-Rx link at the SU-Tx.
- We demonstrate the very high sensitivity of SU performance to the error in the estimation of the PU-Tx to PU-Rx and SU-Tx to PU-Rx links.
- We show that errored knowledge of the PU-Tx to PU-Rx link and SU-Tx to PU-Rx link (if available) is better for SU capacity than a knowledge of the mean link gains.
- By considering a single probabilistic SINR constraint, a unified framework is presented which enables fair comparisons between different types of channel knowledge.

II. SYSTEM MODEL

Consider a CR system (shown in Fig. 1) with the SU-Tx and PU-Tx transmitting simultaneously to their respective receivers. Independent point-to-point flat Rayleigh fading channels are assumed for all links. Let $g_p = |h_p|^2$, $g_s = |h_s|^2$, $g_{ps} = |h_{ps}|^2$ and $g_{sp} = |h_{sp}|^2$ denote the exponentially

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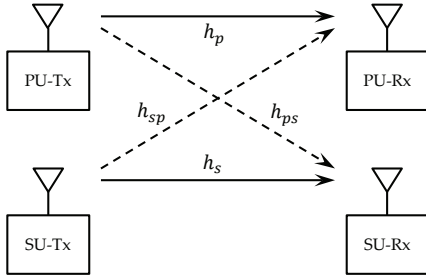


Fig. 1. System Diagram.

distributed instantaneous link gains of the PU-Tx to PU-Rx, SU-Tx to SU-Rx, PU-Tx to SU-Rx and SU-Tx to PU-Rx links, respectively, with $\Omega_p = \mathbb{E}(g_p)$, $\Omega_s = \mathbb{E}(g_s)$, $\Omega_{ps} = \mathbb{E}(g_{ps})$ and $\Omega_{sp} = \mathbb{E}(g_{sp})$, where $\mathbb{E}(\cdot)$ denotes the expectation operator.

As described further in this Section, the SU transmission under the SINR constraint is governed solely by the state of the g_p and g_{sp} links¹. Thus, in this paper we consider the following five scenarios for the knowledge of g_p and g_{sp} by the SU-Tx.

Scenario 1: The link gains, g_p and g_{sp} , are perfectly known. This clearly unrealistic scenario serves as a benchmark for comparison.

Scenario 2: The link gain, g_p , is perfectly known while only the mean, Ω_{sp} , of g_{sp} is known.

Scenario 3: The mean, Ω_p , and the exact link gain, g_{sp} , are known. In contrast to *Scenario 2*, this case is considered mainly for completeness.

Scenario 4: Only the means, Ω_p and Ω_{sp} , are known. This scenario arises when only statistical information about the channels is available to the SU-Tx as a result of limited feedback resources.

Scenario 5: Only estimates of g_p and g_{sp} are available. This may arise due to channel estimation, feedback quantisation and delay.

Where possible, we impose a constraint, γ_T , on the PU-Rx SINR, denoted by γ_p . Hence,

$$\gamma_p = \frac{P_p g_p}{P_s g_{sp} + \sigma_p^2}, \text{ and } \gamma_p \geq \gamma_T, \quad (1)$$

where γ_T is an SINR threshold, P_p is the PU transmit power (assumed to be fixed and known to the SU-Tx in all scenarios), and σ_p^2 is the additive white Gaussian noise (AWGN) variance at the PU-Rx. P_s is a provisional maximum value for the SU transmit power chosen to satisfy the relevant criteria, (1) in this case. The actual SU transmit power, P_t is a function of P_s . For example, if the PU-Rx SNR lies in the region, $P_p g_p / \sigma_p^2 < \gamma_T$, then (1) cannot be satisfied unless $P_s < 0$ and as a result, $P_t = 0$. If the PU SNR is above the SINR threshold γ_T , the SU-Tx will adapt P_s to a maximum level satisfying (1) as determined under the five scenarios, regardless of the gain g_s .

¹The link gains g_s and g_{ps} have an impact on achievable SU capacity, however the level of their knowledge by the SU-Tx does not impact the transmit power.

We also impose a maximum SU transmit power constraint, P_m . Thus, in *Scenario 1*, where the SU-Tx knows g_p , P_t is given by

$$P_t = \begin{cases} 0 & \frac{P_p g_p}{\gamma_T} < \sigma_p^2 \\ \min(P_s, P_m) & \text{otherwise,} \end{cases} \quad (2)$$

where P_s is obtained from (1) by solving $\gamma_T = \gamma_p$. Furthermore, the constraints described above can only be guaranteed if the SU-Tx has perfect knowledge of the gains g_p and g_{sp} , i.e., under *Scenario 1*. In analysing *Scenarios 2-5*, we use probabilistic constraints. Hence, we require the SINR constraint to hold with an acceptably high probability, $1 - \alpha$, where α is small.

In analysing the SU capacity, we first consider the SINR at the SU-Rx, denoted by γ_I ,

$$\gamma_I = \frac{P_t g_s}{P_p g_{ps} + \sigma_s^2}, \quad (3)$$

where σ_s^2 is the AWGN variance at the SU-Rx and $P_p g_{ps}$ is the interference from PU-Tx, treated as noise in the capacity calculations. We denote the pdf and cdf of γ_I by $f_{\gamma_I}(x)$ and $F_{\gamma_I}(x)$, respectively. The instantaneous SU capacity is given by $C = \log_2(1 + \gamma_I)$, where the mean, \bar{C} , can be derived using $f_{\gamma_I}(x)$ as

$$\bar{C} = \mathbb{E}(C) = \int_0^\infty \log_2(1 + x) f_{\gamma_I}(x) dx. \quad (4)$$

The cdf of C can be obtained from $F_{\gamma_I}(x)$ by noting that

$$F_C(y) = \Pr(\gamma_I \leq 2^y - 1) = F_{\gamma_I}(\tilde{y}), \quad (5)$$

where $\Pr(\cdot)$ denotes probability and $\tilde{y} = 2^y - 1$. Using (3), we can express (5) as

$$\begin{aligned} F_{\gamma_I}(\tilde{y}) &= \mathbb{E}_{g_{ps}} \left\{ \Pr(P_t g_s < \tilde{y}(\sigma_s^2 + P_p g_{ps})) \middle| g_{ps} \right\} \\ &= \int_0^\infty F_\gamma(\tilde{y}(\sigma_s^2 + P_p v)) \frac{e^{-v/\Omega_{ps}}}{\Omega_{ps}} dv, \end{aligned} \quad (6)$$

where we have defined $\gamma = P_t g_s$ with a cdf $F_\gamma(x)$. In what follows, we derive expressions for $F_\gamma(x)$ which, using (5) and (6), allows us to compute the capacity cdf.

We parameterize the main system variables by two key parameters. The first, $c_1 = \Omega_{sp}/\Omega_s$, represents the ratio of the mean interference at the PU-Rx to the mean of the desired channel strength for the SU. The second, $c_2 = \gamma_T \sigma_p^2 / P_p \Omega_p$, is the ratio of the minimum target SINR to the mean SNR at the PU-Rx. Hence, increasing c_2 corresponds to reducing the allowable interference (with $c_2 = 1$ corresponding to zero average allowable interference).

III. SU CAPACITY

The capacity mean in (4) and the cdf in (6) require a knowledge of the distributions of $\gamma = P_t g_s$ and γ_I . Hence, in this section we derive the cdfs for γ and γ_I for *Scenarios 1-4*. For *Scenario 5*, an alternative approach is required (see Section III-E).

A. Scenario 1

In this scenario, P_s can be obtained directly from (1), giving

$$P_s = \frac{P_p g_p - \sigma_p^2}{\gamma_T g_{sp}}. \quad (7)$$

We note that while we ignore the $P_t = 0$ case in (2), the following derivation is valid since $\Pr(\gamma > 0) = 0$ for $P_t \leq 0$. In finding $F_\gamma(x)$, we solve for the complementary cdf given by

$$\begin{aligned} \Pr(\gamma > x) &= \Pr(g_s \min(P_m, P_s) > x) \\ &= \Pr\left(g_s > \frac{x}{P_m}, \left(\frac{P_p g_p}{\gamma_T} - \sigma_p^2\right) g_s > x g_{sp}\right). \end{aligned} \quad (8)$$

Noting that g_p is an exponentially distributed RV, we can rewrite (8) by taking the conditional probability over g_p and then averaging over g_s and g_{sp} . This gives,

$$\Pr(\gamma > x) = \frac{e^{-\frac{\gamma_T \sigma_p^2}{P_p \Omega_p}}}{\Omega_{sp} \Omega_s} \int_{\frac{x}{P_m}}^{\infty} \frac{e^{-\frac{u}{\Omega_s}}}{\frac{\gamma_T x}{P_p \Omega_p u} + \frac{1}{\Omega_{sp}}} du. \quad (9)$$

After simplifying (9), the cdf $F_\gamma(x) = 1 - \Pr(\gamma > x)$ can be shown to be [15, Eq. (3.351.2)]

$$\begin{aligned} F_\gamma(x) &= 1 - e^{-\frac{\gamma_T \sigma_p^2}{P_p \Omega_p}} \left[e^{-\frac{x}{P_m \Omega_s}} - \frac{\Omega_{sp} \gamma_T x}{P_p \Omega_p \Omega_s} e^{\frac{\Omega_{sp} \gamma_T x}{P_p \Omega_p \Omega_s}} \right. \\ &\quad \left. \times \Gamma\left(0, \frac{\Omega_{sp} \gamma_T x}{P_p \Omega_p \Omega_s} + \frac{x}{P_m \Omega_s}\right) \right], \end{aligned} \quad (10)$$

where $\Gamma(\cdot, \cdot)$ is the upper incomplete gamma function. Substituting (10) into (6) results in

$$\begin{aligned} F_{\gamma_I}(\tilde{y}) &= 1 - \frac{P_m \Omega_s e^{-\left(\frac{\gamma_T \sigma_p^2}{P_p \Omega_p} + \frac{\tilde{y} \sigma_s^2}{P_m \Omega_s}\right)}}{P_m \Omega_s + \tilde{y} P_p \Omega_{ps}} \\ &\quad + \frac{\Omega_{sp} \gamma_T \tilde{y}}{\Omega_{ps} \Omega_p \Omega_s P_p} \exp\left\{\frac{\Omega_{sp} \gamma_T \sigma_p^2}{\Omega_p \Omega_s P_p} \left(\tilde{y} - \frac{\Omega_s}{\Omega_{sp}}\right)\right\} \\ &\quad \times \int_0^\infty (\sigma_p^2 + P_p v) \exp\left\{\left(\frac{\Omega_{sp} \gamma_T}{\Omega_p \Omega_s} \tilde{y} - \frac{1}{\Omega_{ps}}\right) v\right\} \\ &\quad \times \Gamma\left(0, \frac{\Omega_{sp} \gamma_T P_m + P_p \Omega_p}{P_p P_m \Omega_p \Omega_s} (\sigma_p^2 + P_p v) \tilde{y}\right) dv. \end{aligned} \quad (11)$$

To the best of the authors' knowledge, the integral in (11) does not have a closed-form solution. In Section IV, the capacity cdf results are obtained by numerical integration.

In order to obtain the expression for mean capacity, we can derive the pdf, $f_{\gamma_I}(x)$, by differentiating (11) with respect to \tilde{y} . Alternatively, using (6) we have

$$f_{\gamma_I}(\tilde{y}) = \int_0^\infty (\sigma_p^2 + P_p v) f_\gamma(\tilde{y}(\sigma_p^2 + P_p v)) \frac{e^{-v/\Omega_{ps}}}{\Omega_{ps}} dv, \quad (12)$$

where $f_\gamma(x)$ was computed in [16] as,

$$\begin{aligned} f_\gamma(x) &= e^{-\frac{\gamma_T \sigma_p^2}{P_p \Omega_p}} \left[\left(\frac{1}{P_m \Omega_s} - \frac{\Omega_{sp} \gamma_T}{P_p \Omega_p \Omega_s} \right) e^{-\frac{x}{P_m \Omega_s}} \right. \\ &\quad \left. + e^{\frac{\Omega_{sp} \gamma_T x}{P_p \Omega_p \Omega_s}} \left(\frac{(\Omega_{sp} \gamma_T)^2 x}{(P_p \Omega_p \Omega_s)^2} + \frac{\Omega_{sp} \gamma_T}{P_p \Omega_p \Omega_s} \right) \right. \\ &\quad \left. \times \Gamma\left(0, \frac{\Omega_{sp} \gamma_T x}{P_p \Omega_p \Omega_s} + \frac{x}{P_m \Omega_s}\right) \right]. \end{aligned} \quad (13)$$

The expression resulting from substituting (13) into (12) cannot be written in closed-form. Thus, the mean capacity, \bar{C} , must be calculated numerically by substituting (13) into (12) and (4).

B. Scenario 2

In *Scenarios 2-5*, with imperfect channel knowledge, the SU cannot guarantee that (1) is satisfied. Thus, we constrain the SU to satisfy (1) with an acceptably high probability, $1 - \alpha$.

Hence, for *Scenario 2*, where the SU knows only the mean, Ω_{sp} , of g_{sp} , we consider the probability of satisfying the SINR constraint with a probability of $1 - \alpha$. This gives

$$\Pr\left(\frac{P_p g_p}{P_s g_{sp} + \sigma_p^2} \geq \gamma_T \mid g_p, \Omega_{sp}\right) = 1 - \alpha. \quad (14)$$

Rewriting in terms of the cdf of g_{sp} , we derive P_s as

$$P_s = -\frac{P_p g_p - \gamma_T \sigma_p^2}{\ln(\alpha) \gamma_T \Omega_{sp}}. \quad (15)$$

Using (15), the complementary cdf of γ can be shown to be

$$\Pr(\gamma > x) = \mathbb{E}\left[\Pr\left(P_m g_s > x, P_s g_s > x \mid g_p\right)\right], \quad (16)$$

which can be expressed as

$$\begin{aligned} \Pr(\gamma > x) &= \int_{\psi_0}^{\psi} \Pr\left(g_s > \frac{x}{P_s}\right) f_{g_p}(y) dy \\ &\quad + \int_{\psi}^{\infty} \Pr\left(g_s > \frac{x}{P_m}\right) f_{g_p}(y) dy, \end{aligned} \quad (17)$$

where $\psi_0 = \frac{\gamma_T \sigma_p^2}{P_p}$ and $\psi = \frac{\gamma_T (\sigma_p^2 - P_m \ln(\alpha) \Omega_{sp})}{P_p}$. The lower integration limit in the first term of (17) takes into account the $P_t = 0$ condition in (2). After some manipulation, we can simplify (17) to obtain $F_\gamma(x) = 1 - \Pr(\gamma > x)$ as,

$$\begin{aligned} F_\gamma(x) &= 1 - \exp\left\{-\frac{x}{P_m \Omega_s} - \frac{\psi}{\Omega_p}\right\} \\ &\quad - \frac{1}{\Omega_p} \int_{\psi_0}^{\psi} e^{-\frac{\ln(\alpha) \gamma_T \Omega_{sp} x}{(\gamma_T \sigma_p^2 - P_p y) \Omega_s}} e^{-\frac{y}{\Omega_p}} dy. \end{aligned} \quad (18)$$

Once again, there exists no closed-form solution to the integral in (18). Following the same approach as in *Scenario 1*, we use

(18) and (6) to find $F_{\gamma_I}(\tilde{y})$ as

$$F_{\gamma_I}(\tilde{y}) = 1 - \frac{P_m \Omega_s e^{-\left(\frac{\tilde{y} \sigma_p^2}{P_m \Omega_s} + \frac{\psi}{\Omega_p}\right)}}{\Omega_{ps} P_p \tilde{y} + P_m \Omega_s} \quad (19)$$

$$+ \frac{1}{\Omega_s} \int_{\psi_0}^{\psi} e^{-\left(\frac{\Omega_{sp} \gamma_T \sigma_p^2 \ln(\alpha) \tilde{y}}{\gamma_T \sigma_p^2 \Omega_s - P_p \Omega_s z} + \frac{z}{\Omega_s}\right)}$$

$$\times \frac{\gamma_T \sigma_p^2 \Omega_s - P_p \Omega_s z}{\gamma_T \sigma_p^2 \Omega_s + \Omega_{sp} \gamma_T P_p \Omega_{ps} \ln(\alpha) \tilde{y} - P_p \Omega_s z} dz.$$

Here, again, the capacity cdf is obtained using (5) and numerically integrating (19).

To compute the SU mean capacity, we differentiate (19) with respect to \tilde{y} to find the pdf

$$f_{\gamma_I}(x) = -\sigma_p^2 e^{-\left(\frac{x \sigma_p^2}{P_m \Omega_s} + \frac{\psi}{\Omega_p}\right)} \quad (20)$$

$$+ \frac{\Omega_{sp} \gamma_T \sigma_p^2 \ln(\alpha)}{\Omega_s} \int_{\psi_0}^{\psi} e^{-\left(\frac{\Omega_{sp} \gamma_T \sigma_p^2 \ln(\alpha) x}{\gamma_T \sigma_p^2 \Omega_s - P_p \Omega_s z} + \frac{z}{\Omega_s}\right)}$$

$$\times (\gamma_T \sigma_p^2 \Omega_s + \Omega_{sp} \gamma_T P_p \Omega_{ps} \ln(\alpha) x - P_p \Omega_s z)^{-2}$$

$$\times ((\gamma_T \sigma_p^2 \Omega_s - P_p \Omega_s z)(\Omega_{sp} \gamma_T P_p \Omega_{ps} \ln \alpha - 1)$$

$$+ \Omega_{sp} \gamma_T P_p \Omega_{ps} \ln(\alpha) x) dz.$$

The mean capacity is then computed by substituting (20) into (4) and numerically integrating.

C. Scenario 3

In *Scenario 3*, where the SU has exact knowledge of g_{sp} and knows only the mean Ω_p , we once again satisfy the SINR constraint with a probability of $1 - \alpha$. Hence,

$$\Pr\left(\frac{P_p g_p}{P_s g_{sp} + \sigma_p^2} \geq \gamma_T \middle| \Omega_p, g_{sp}\right) = 1 - \alpha. \quad (21)$$

Following the same approach as for *Scenario 2*, one can show that

$$P_s = -\left(\frac{\ln(1 - \alpha) P_p \Omega_p}{\gamma_T} + \sigma_p^2\right) \frac{1}{g_{sp}}. \quad (22)$$

From (22), the SU SINR cdf, $F_{\gamma_I}(\tilde{y})$, can be derived as (see [17, Appendix A] for details)

$$F_{\gamma_I}(\tilde{y}) = 1 - s(\tilde{y}) - h(\tilde{y})E_1(r(\tilde{y})), \quad (23)$$

where $s(\tilde{y})$, $h(\tilde{y})$ and $r(\tilde{y})$ are given by

$$s(\tilde{y}) = \frac{K_1 e^{-b\tilde{y}}}{1 + a\tilde{y}}, \quad h(\tilde{y}) = \frac{K_2 e^{-b\tilde{y} + r(\tilde{y})}}{\tilde{y}}, \quad (24)$$

$$r(\tilde{y}) = \frac{(P_p \Omega_{ps} \tilde{y} + P_m \Omega_s)(\sigma_s^2 \Omega_{sp} \tilde{y} + Q \Omega_s)}{P_m P_p \Omega_s \Omega_{ps} \Omega_{sp} \tilde{y}},$$

with constants, $K_1 = 1 - e^{Q/P_m \Omega_{sp}}$, $K_2 = \frac{Q \Omega_s e^{Q/P_m \Omega_{sp}}}{P_p \Omega_{ps} \Omega_{sp}}$, $a = \frac{P_p \Omega_{ps}}{P_m \Omega_s}$ and $b = \sigma_s^2 / P_m \Omega_s$. Hence,

$$f_{\gamma_I}(\tilde{y}) = -s'(\tilde{y}) - h'(\tilde{y})E_1(r(\tilde{y})) + h(\tilde{y})r'(\tilde{y}) \frac{e^{-r(\tilde{y})}}{r(\tilde{y})}, \quad (25)$$

where $s'(\tilde{y})$, $h'(\tilde{y})$ and $r'(\tilde{y})$ are the derivatives of (24).

D. Scenario 4

When the SU-Tx has knowledge of only Ω_p and Ω_{sp} , then we have

$$\Pr\left(\frac{P_p g_p}{P_s g_{sp} + \sigma_p^2} \geq \gamma_T \middle| \Omega_p, \Omega_{sp}\right) = 1 - \alpha. \quad (26)$$

Using conditioning and after some manipulation, P_s can be derived as

$$P_s = \frac{P_p \Omega_p}{\gamma_T \Omega_{sp}} \left(\frac{e^{-\frac{\gamma_T \sigma_p^2}{P_p \Omega_p}}}{1 - \alpha} - 1 \right). \quad (27)$$

Here, P_s and P_t are deterministic, depending on the system parameters. The latter is given by

$$P_t = \begin{cases} 0 & P_s < 0 \\ P_s & 0 < P_s < P_m \\ P_m & P_s > P_m. \end{cases} \quad (28)$$

Using (28), we obtain the cdf of $\gamma = P_t g_s$, which, when substituted into (6) and (5), results in

$$F_C(y) = 1 - \frac{P_t \Omega_s}{\tilde{y} P_p \Omega_{ps} + P_t \Omega_s} e^{-\frac{\tilde{y} \sigma_s^2}{P_t \Omega_s}}. \quad (29)$$

In order to compute the mean capacity, \bar{C} , we note that $F_{\gamma_I}(x)$ can be trivially derived from (29) and (5). Differentiating to obtain $f_{\gamma_I}(x)$ and substituting into (4), one obtains

$$\bar{C} = \frac{1}{\ln(2)} \int_1^\infty \left(\frac{\sigma_s^2}{P_p \Omega_{ps} t + P_t \Omega_s} + \frac{P_t P_p \Omega_s \Omega_{ps}}{(P_p \Omega_{ps} t + P_t \Omega_s)^2} \right)$$

$$\times \ln(t) e^{-\frac{t \sigma_s^2}{P_t \Omega_s}} dt, \quad (30)$$

where we have used the change of variable, $t = 1 + x$.

E. Scenario 5

Here, the SU-Tx operates on estimates of g_p and g_{sp} . In such a case, we aim to satisfy:

$$\Pr\left(P_p g_p \geq \gamma_T P_s g_{sp} + \gamma_T \sigma_s^2 \middle| \hat{g}_p, \hat{g}_{sp}\right) = 1 - \alpha, \quad (31)$$

which must be solved for P_s . We use the classic model for imperfect CSI [9] given by $\hat{h} = \rho h + \sqrt{1 - \rho^2} \epsilon$, where h is a generic channel coefficient, ρ controls the level of CSI, ϵ is statistically identical to h and $\hat{g} = |\hat{h}|^2$. The complexity of (31) makes further capacity analysis infeasible. Instead, (31) is derived in [17, Appendix B] and is shown to be equivalent to

$$\sum_{j=0}^{\infty} \frac{(\lambda_1/2)^j}{j!} e^{-\lambda_1/2} \left(1 - e^{-\frac{\lambda_2 + \beta}{2}} e^{\frac{\lambda_2}{4(\delta+1)}} \sqrt{\frac{8}{\lambda_2(\delta+1)}} \right) \quad (32)$$

$$\times \sum_{r=0}^j \sum_{s=0}^r \left(\frac{\beta}{2} \right)^r \frac{\left(\frac{2\delta}{\beta(\delta+1)} \right)^s}{(r-s)!} M_{-s-1/2,0} \left(\frac{\lambda_2}{2(\delta+1)} \right) = \alpha,$$

where $\lambda_1 = 2\rho^2 \hat{g}_p / (\Omega_p (1 - \rho^2))$, $\lambda_2 = 2\rho^2 \hat{g}_{sp} / (\Omega_{sp} (1 - \rho^2))$, $\beta = 2\sigma_s^2 \gamma_T / (\Omega_p (1 - \rho^2) P_p)$, $\delta = \gamma_T P_s \Omega_{sp} / (\Omega_p P_p)$ and $M_{\mu,\nu}(\cdot)$ is a Whittaker function. P_s is then computed using a numerical root finder to solve (32) and the resulting value is used in capacity simulations.

F. SU Blocking

Using the results in Sections III-A - III-D, we derive the SU blocking conditions, that is the probability or condition under which the SU cannot transmit due to the constraint (1).

In the case of *Scenarios* 1 and 2, where P_s is dependent on the instantaneous value of g_p , via (7) and (15), respectively, we can compute the probability of SU blocking, by solving for $\Pr(P_t \leq 0)$ or equivalently $\Pr(P_s \leq 0)$. It is easily shown that for *Scenarios* 1 and 2

$$\Pr(P_t \leq 0) = 1 - e^{-\frac{\gamma_T \sigma_p^2}{\Omega_p P_p}} = 1 - e^{-c_2}. \quad (33)$$

For *Scenarios* 3 and 4, the SU blocking condition is determined purely from the system parameters, and can be obtained by setting $P_s \leq 0$ in (22) and (27), respectively. Here, the SU blocking condition is related to α and c_2 by

$$P_t = 0 \quad \text{if} \quad \alpha \leq 1 - e^{-\frac{\gamma_T \Omega_p}{\sigma_p^2 P_p}} = 1 - e^{-c_2}. \quad (34)$$

Using (34), we note that for small values of α , that is where we guarantee the PU SINR constraint with high probability, the SU blocking condition is approximated by $\alpha \leq c_2$.

For *Scenario* 5, blocking occurs when (31) can not be satisfied, even for $P_s = 0$. Hence, the boundary of the blocking region is equivalent to

$$\Pr \left\{ \Pr \left(g_p \geq \frac{\gamma_T \sigma_p^2}{P_p} \middle| \hat{g}_p \right) = 1 - \alpha \right\}. \quad (35)$$

In [17], the probability in (35) is rewritten as

$$\Pr \left(X \geq \frac{2c_2}{1 - \rho^2} \middle| \hat{g}_p \right) = 1 - \alpha, \quad (36)$$

where X is a non-central Chi-squared variable with 2 degrees of freedom and non-centrality parameter, $2\rho^2 \hat{g}_p / (\Omega_p (1 - \rho^2))$. We solve (36) by a simple root-finder, to find the threshold value, $\hat{g}_p = g^*$, which satisfies (36). Then, the blocking probability is simply

$$\Pr(\hat{g}_p < g^*) = 1 - e^{-g^*/\Omega_p}. \quad (37)$$

IV. SIMULATION RESULTS AND DISCUSSION

In all simulations, we have set $P_p/\sigma_p^2 = P_m/\sigma_s^2 = 0$ dB and $\Omega_p/\sigma_p^2 = \Omega_s/\sigma_s^2 = 5$ dB, where we assume $\sigma_p^2 = \sigma_s^2$. In *Scenarios* 2-5 we set $\alpha = 0.1$, and $\rho = 0.9$ is used in *Scenario* 5, unless otherwise indicated in the figures. Figures 2, 3 and 5 show the SU capacity cdfs for various scenarios and a range of c_1, c_2 values. Figure 5, with $c_1 = 0.01$, represents very favourable SU operating conditions. Figure 3 ($c_1 = 0.1, c_1 = 0.9$) represents increasingly difficult conditions for the SU.

From these results, we observe that *Scenarios* 1 and 2 result in similar performance, even in the case of $c_1 = 0.9$ (Fig. 3), that is where the SU interference is very prominent. Furthermore, lack of knowledge of the PU-PU link (knowing only the mean Ω_p) greatly reduces the achievable capacity of the SU. This is shown in Figs. 2 and 3 where *Scenarios* 3 and 4 suffer a considerable loss in comparison to *Scenarios* 1 and 2. Hence, knowledge of g_p is more important than g_{sp} .

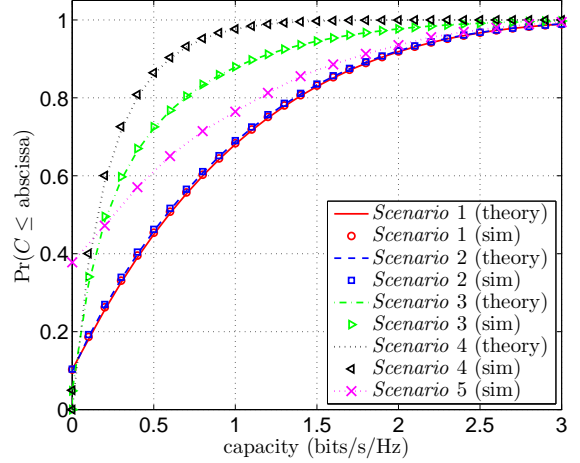


Fig. 2. SU capacity cdf for *Scenarios* 1-5 ($c_1 = c_2 = 0.1$).

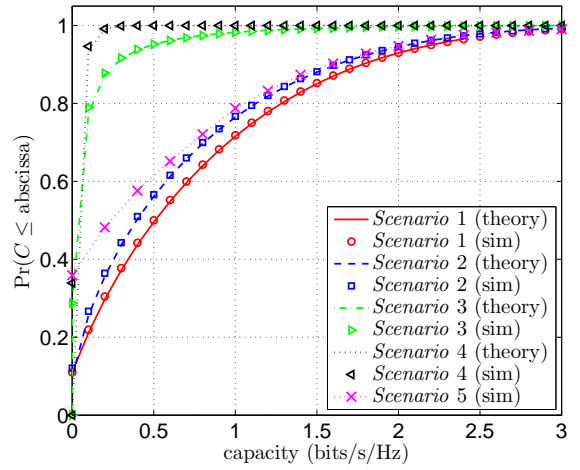


Fig. 3. SU capacity cdf for *Scenarios* 1-5 ($c_1 = 0.9, c_2 = 0.1$).

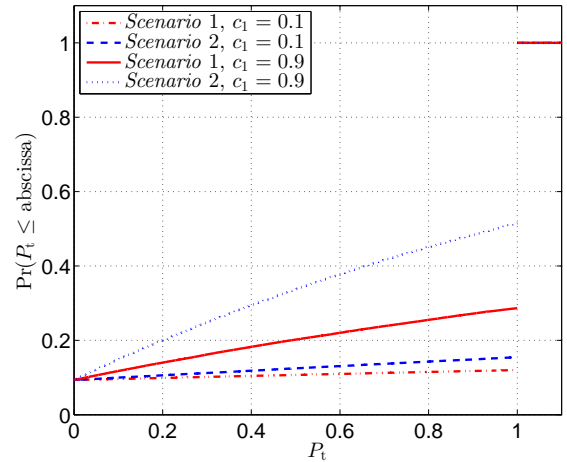


Fig. 4. SU transmit power cdf for *Scenarios* 1 and 2 ($c_1 = 0.1$ and $c_1 = 0.9$).

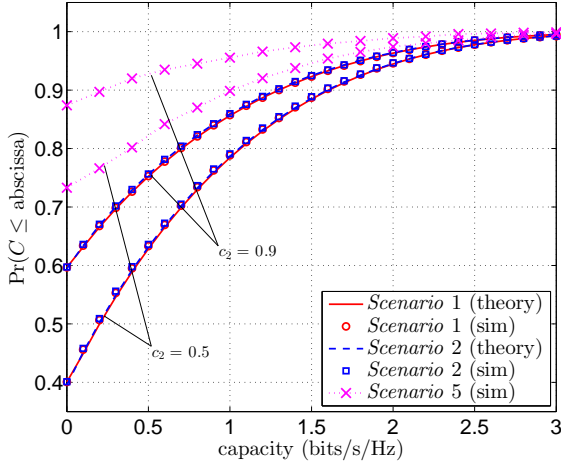


Fig. 5. SU capacity cdf for *Scenarios* 1, 2 and 5 ($c_1 = 0.01$; $c_2 = 0.5$, $c_2 = 0.9$)

The dependence on c_1 can be observed by comparing Figs. 2 and 3. Under very favourable conditions, $c_1 = 0.01$, *Scenarios* 3 and 4 slightly outperform *Scenarios* 1 and 2. This seemingly counterintuitive result is due to the flexibility afforded by the probabilistic SINR constraint. This is confirmed by the additional cdfs for *Scenarios* 3 and 4 in Fig. 3, with $\alpha = 0.096$, where the protection of PU SINR with higher degree of certainty causes degradation of performance for *Scenarios* 3 and 4 below that for *Scenarios* 1 and 2.

From Fig. 3, we observe that placing the SU in a demanding environment, $c_1 = 0.9$, results in very poor performance under *Scenarios* 3 and 4. Furthermore, the performance of *Scenario* 2 is noticeably degraded from that of *Scenario* 1. Further insight into this is provided by Fig. 4, which shows the cdf of the SU transmit power, P_t , for $c_1 = 0.1$ and $c_1 = 0.9$. We observe that in the latter case, the SU-Tx under *Scenario* 1 operates at maximum power, $P_t = 1$, with a likelihood of 70 %, compared to approximately 50 % for *Scenario* 2. This difference is much less pronounced for the less challenging case of $c_1 = 0.1$. Finally, based on Figs. 2 and 3, we observe that the performance under *Scenario* 5 is not highly dependent on the value of c_1 .

Comparing the curves for *Scenarios* 3 and 4 with those for *Scenario* 5 in Fig. 3, we note that, for the most part, imperfect knowledge of the link gains is more beneficial than a knowledge of their means. Only in the low capacity regime we observe that *Scenarios* 3 and 4 outperform *Scenario* 5, which has a relatively high blocking probability for the parameters considered. It should be noted, however, that blocking in *Scenarios* 3 and 4 is dictated by the parameter c_2 and thus, unless (34) is satisfied, the capacity cdfs for these scenarios originate at zero. Consequently, at higher capacity values there exists a crossover point with *Scenario* 5.

Figures 2 and 3 compare the scenarios using $c_2 = 0.1$, which is very generous to the SU. From (34), we see that SU transmission in *Scenarios* 3 and 4 occurs only for large values of α or for small values of c_2 . That is, without the knowledge of g_p , the SU can only operate if the PU is willing to accept

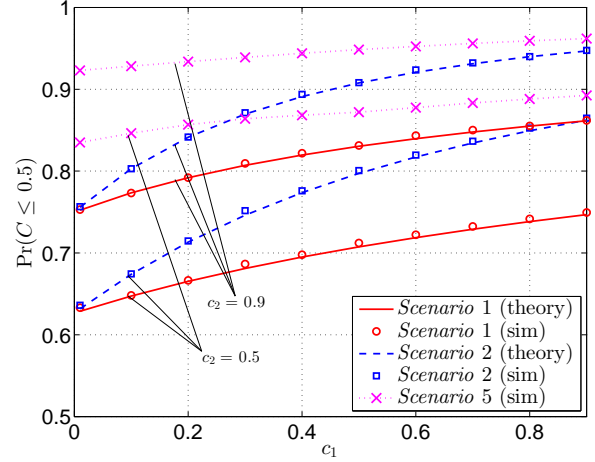


Fig. 6. SU capacity outage values for *Scenarios* 1, 2 and 5 vs c_1 ($c_2 = 0.5$, $c_2 = 0.9$).

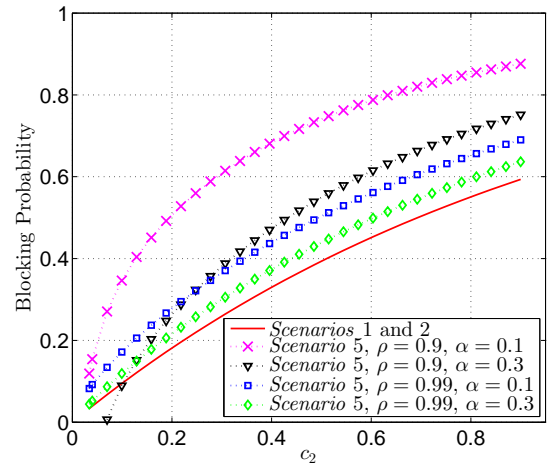


Fig. 7. Blocking probability for *Scenarios* 1, 2 and 5 for vs c_2 .

large amounts of interference. Figure 5 presents the capacity results for *Scenarios* 1 and 2 with the more realistic values of $c_2 = 0.5$ and $c_2 = 0.9$, where (34) prevents SU transmission under *Scenarios* 3 and 4. While SU transmission is possible under *Scenario* 5, we observe a high blocking probability of 0.73 and 0.88 for $c_2 = 0.5$ and $c_2 = 0.9$, respectively.

Figure 6 shows the probability $\Pr(C \leq 0.5)$ as a function of c_1 . As expected, for a constant c_2 , the performance under *Scenario* 2 diverges from the baseline *Scenario* 1 with increasing c_1 , that is as the amount of interference to the PU increases.

Finally, Fig. 7 shows the blocking probability for *Scenarios* 1, 2 and 5. We recall that the SU ability to transmit in *Scenarios* 3 and 4 is deterministic and governed by the blocking condition of (34). The results for *Scenario* 5 were obtained numerically via (36). We observe that as the channel knowledge error decreases ($\rho \rightarrow 1$), the blocking probability approaches that of *Scenarios* 1 and 2. Specifically, referring back to Fig. 5, we note from Fig. 7 where $\rho = 0.9$ that improving the channel estimate to $\rho = 0.99$ will reduce

the blocking probability at $c_2 = 0.5$ and $c_2 = 0.9$ to 0.5 and 0.7, respectively, thus bringing the performance of *Scenario 5* closer to that of *Scenario 1*. Similarly, relaxing the probabilistic SINR constraint by increasing α to 0.3 results in a significant reduction in blocking probability, as fully expected.

V. CONCLUSIONS

We have examined the effects of limited channel knowledge on the SU capacity. Considering five scenarios, we derived (in four cases) analytical expressions for the SU capacity cdf under a PU-Rx SINR constraint. We determined the SU blocking probability and blocking conditions as a function of the allowable interference at the PU-Rx. The results demonstrate the importance of the PU-PU CSI, which was shown to be much greater than that of the SU-Tx to PU-Rx link. Furthermore, we have shown that in challenging situations or in the presence of CSI error there can be extremely large blocking probabilities for the SUs.

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