

Strategic Sensing in Cognitive Radio Networks

Ran Snitkovsky

Under Supervision of
Prof. Refael Hassin

The Raymond and Beverly Sackler Faculty of Exact Sciences
The Department of Statistics and Operations Research
Tel-Aviv University

September 26, 2015

Introduction

The problem:

Introduction

The problem:

The problem:

THE RADIO SPECTRUM

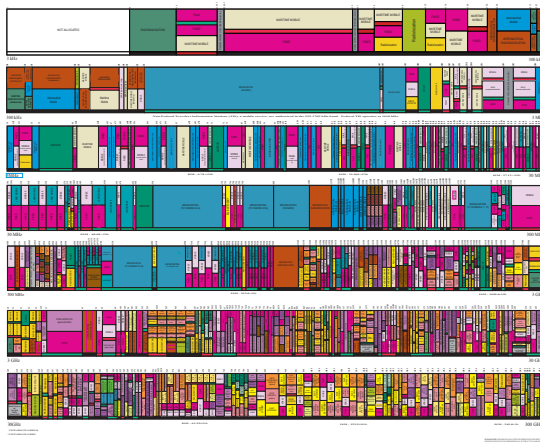


Figure: US frequency allocations of the radio spectrum

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▷ Radio-scene analysis.
 - ▷ Channel-state estimation and predictive modeling.
 - ▷ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▶ Radio-scene analysis.
 - ▶ Channel-state estimation and predictive modeling.
 - ▶ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▶ Radio-scene analysis.
 - ▶ Channel-state estimation and predictive modeling.
 - ▶ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▶ Radio-scene analysis.
 - ▶ Channel-state estimation and predictive modeling.
 - ▶ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▷ Radio-scene analysis.
 - ▷ Channel-state estimation and predictive modeling.
 - ▷ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▷ Radio-scene analysis.
 - ▷ Channel-state estimation and predictive modeling.
 - ▷ **Transmit-power control and dynamic spectrum management.**

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▷ Radio-scene analysis.
 - ▷ Channel-state estimation and predictive modeling.
 - ▷ Transmit-power control and dynamic spectrum management.

An optional solution:

Cognitive Radio (CR) - A dynamically configured transceiver.

- ▶ First published by Mitola, J. & Maguire, G.Q., Jr. (1999).
- ▶ Haykin, S. (2005) addresses three fundamental cognitive tasks:
 - ▷ Radio-scene analysis.
 - ▷ Channel-state estimation and predictive modeling.
 - ▷ **Transmit-power control and dynamic spectrum management.**

- ▶ Many researches have associated CR with “Rational Queueing” .
- ▶ Do, C. T. *et al.* (2012) solve a generalization of Edelson & Hildebrand’s Unobservable Model with server’s breakdowns for spectrum sharing in CRN.
- ▶ Habachi, O. & Hayel, Y. (2012) discuss a rather general model of opportunistic sensing, but with no regard to failed-sensing consequences.

- ▶ Many researches have associated CR with “Rational Queueing”.
- ▶ Do, C. T. *et al.* (2012) solve a generalization of Edelson & Hildebrand’s Unobservable Model with server’s breakdowns for spectrum sharing in CRN.
- ▶ Habachi, O. & Hayel, Y. (2012) discuss a rather general model of opportunistic sensing, but with no regard to failed-sensing consequences.

- ▶ Many researches have associated CR with “Rational Queueing”.
- ▶ Do, C. T. *et al.* (2012) solve a generalization of Edelson & Hildebrand’s Unobservable Model with server’s breakdowns for spectrum sharing in CRN.
- ▶ Habachi, O. & Hayel, Y. (2012) discuss a rather general model of opportunistic sensing, but with no regard to failed-sensing consequences.

- ▶ Many researches have associated CR with “Rational Queueing”.
- ▶ Do, C. T. *et al.* (2012) solve a generalization of Edelson & Hildebrand’s Unobservable Model with server’s breakdowns for spectrum sharing in CRN.
- ▶ Habachi, O. & Hayel, Y. (2012) discuss a rather general model of opportunistic sensing, but with no regard to failed-sensing consequences.

System-Model

Two identical servers, S_Q and S_L , and a single queue.

Customers choose between two options: *Sense* S_L or *Join* S_Q .

System-Model

Two identical servers, S_Q and S_L , and a single queue.

Customers choose between two options: *Sense* S_L or *Join* S_Q .

System-Model

Two identical servers, S_Q and S_L , and a single queue.

Customers choose between two options: *Sense* S_L or *Join* S_Q .

System-Model

Two identical servers, S_Q and S_L , and a single queue.

Customers choose between two options: *Sense* S_L or *Join* S_Q .

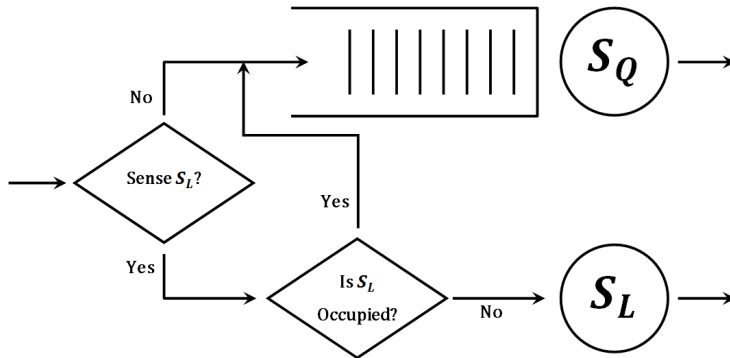


Figure: Customers' flow chart of the system

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

Formulation:

- ▶ Identical rational individualistic customers
- ▶ Arrival rate $\sim \text{Poisson}(\Lambda)$
- ▶ Service duration $\sim \text{Exp}(\mu)$
- ▶ p - sensing probability
- ▶ $(X(t), Y(t))$ - the state at time t , where $X(t) \in \{0, 1, 2, \dots\}$ and $Y(t) \in \{0, 1\}$
- ▶ $P_{i,j}$ - the stationary probability of state (i, j)

System-Model

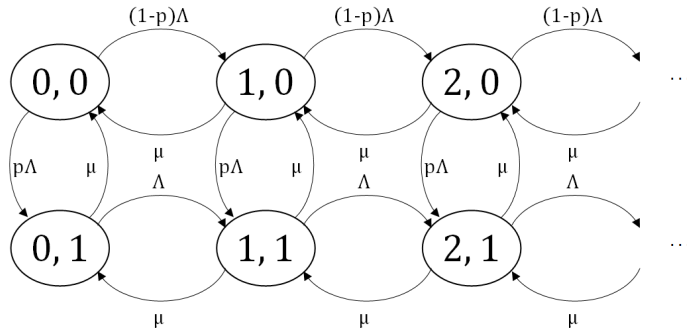


Figure: The Markov chain describing the transitions between states in the system

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

This is a particular case of the *Heterogeneous Arrivals & Service* model of Yechiali, U. & Naor, P. (1971).

S_L is an independent M/M/1/1

The stationary probabilities are found solving the set of equations:

$$\begin{cases} \Lambda P_{0,0} - \mu P_{1,0} - \mu P_{0,1} = 0, \\ (\mu + \Lambda) P_{0,1} - p \Lambda P_{0,0} - \mu P_{1,1} = 0. \end{cases} \quad (1)$$

and $\forall n \in \{1, 2, \dots\}$:

$$\begin{cases} (\mu + \Lambda) P_{n,0} - (1 - p) \Lambda P_{n-1,0} - \mu P_{n+1,0} - \mu P_{n,1} = 0, \\ (2\mu + \Lambda) P_{n,1} - p \Lambda P_{n,0} - \Lambda P_{n-1,1} - \mu P_{n+1,1} = 0. \end{cases} \quad (2)$$

Defining $\rho := \Lambda/\mu$, after some development we achieve the following results:

$$\Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,1} = \frac{p\rho}{1 + p\rho} ; \quad (3)$$

$$\Pr(Y = 0) = 1 - \Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,0} = \frac{1}{1 + p\rho} ; \quad (4)$$

$$\begin{aligned} \hat{\rho}(p, \rho) &:= \frac{1}{\mu} [(1 - p)\Lambda + \Pr(Y = 1) \cdot p\Lambda] \\ &= \rho - \frac{1}{1 + p\rho} p\rho = 1 - (P_{0,0} + P_{0,1}) . \end{aligned} \quad (5)$$

Defining $\rho := \Lambda/\mu$, after some development we achieve the following results:

$$\Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,1} = \frac{p\rho}{1 + p\rho} ; \quad (3)$$

$$\Pr(Y = 0) = 1 - \Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,0} = \frac{1}{1 + p\rho} ; \quad (4)$$

$$\begin{aligned} \hat{\rho}(p, \rho) &:= \frac{1}{\mu} [(1 - p)\Lambda + \Pr(Y = 1) \cdot p\Lambda] \\ &= \rho - \frac{1}{1 + p\rho} p\rho = 1 - (P_{0,0} + P_{0,1}) . \end{aligned} \quad (5)$$

Defining $\rho := \Lambda/\mu$, after some development we achieve the following results:

$$\Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,1} = \frac{p\rho}{1 + p\rho} ; \quad (3)$$

$$\Pr(Y = 0) = 1 - \Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,0} = \frac{1}{1 + p\rho} ; \quad (4)$$

$$\begin{aligned} \hat{\rho}(p, \rho) &:= \frac{1}{\mu} [(1 - p)\Lambda + \Pr(Y = 1) \cdot p\Lambda] \\ &= \rho - \frac{1}{1 + p\rho} p\rho = 1 - (P_{0,0} + P_{0,1}) . \end{aligned} \quad (5)$$

Defining $\rho := \Lambda/\mu$, after some development we achieve the following results:

$$\Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,1} = \frac{p\rho}{1 + p\rho} ; \quad (3)$$

$$\Pr(Y = 0) = 1 - \Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,0} = \frac{1}{1 + p\rho} ; \quad (4)$$

$$\begin{aligned} \hat{\rho}(p, \rho) &:= \frac{1}{\mu} [(1 - p)\Lambda + \Pr(Y = 1) \cdot p\Lambda] \\ &= \rho - \frac{1}{1 + p\rho} p\rho = 1 - (P_{0,0} + P_{0,1}) . \end{aligned} \quad (5)$$

Defining $\rho := \Lambda/\mu$, after some development we achieve the following results:

$$\Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,1} = \frac{p\rho}{1 + p\rho} ; \quad (3)$$

$$\Pr(Y = 0) = 1 - \Pr(Y = 1) = \sum_{i=0}^{\infty} P_{i,0} = \frac{1}{1 + p\rho} ; \quad (4)$$

$$\begin{aligned} \hat{\rho}(p, \rho) &:= \frac{1}{\mu} [(1 - p)\Lambda + \Pr(Y = 1) \cdot p\Lambda] \\ &= \rho - \frac{1}{1 + p\rho} p\rho = 1 - (P_{0,0} + P_{0,1}) . \end{aligned} \quad (5)$$

Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



Proposition

For each $\rho \in (0, \varphi)$, there exists a lower bound for p , denoted \underline{p} , such that the system is stable iff $p \in (\underline{p}, 1]$.

Proof.

For stability we demand $\hat{\rho}(p, \rho) < 1$.

Using (5) and isolating p in the inequality we get:

$$p > \frac{\rho - 1}{\rho(2 - \rho)} =: \underline{p}$$

By the monotonicity of $\hat{\rho}(p, \rho)$, we assume $p = 1$ concluding that

$$\hat{\rho}(1, \rho) = \frac{\rho^2}{1 + \rho} < 1 \quad \Leftrightarrow \quad \rho < \frac{1 + \sqrt{5}}{2} = \varphi$$



We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

We assume $\rho < \varphi$ and define:

- ▶ $c_w > 0$ - the waiting cost per unit time.
- ▶ $c_s > 0$ - the cost of sensing.

We denote $\gamma := c_w / \mu c_s$ and write down the costs:

$$\begin{cases} C_N(p) = \frac{c_w}{\mu} \mathbb{E}[L(p)] ; \\ C_S(p) = c_s + \Pr(Y = 1) \cdot \frac{c_w}{\mu} \mathbb{E}[L(p) \mid Y = 1] ; \end{cases} \quad (6)$$

$$\Leftrightarrow \begin{cases} \frac{1}{c_s} C_N(p) = \gamma \mathbb{E}[L(p)] ; \\ \frac{1}{c_s} C_S(p) = 1 + \Pr(Y = 1) \cdot \gamma \mathbb{E}[L(p) \mid Y = 1] . \end{cases} \quad (7)$$

Equilibrium Strategy

Proposition

For every $\rho \in (0, \varphi)$, and for every value $\gamma > 0$, a unique equilibrium strategy $p_e \in [0, 1]$ exists.

This result arises in spite of the fact that the function $C_S(p)$ is not necessarily monotone:

Equilibrium Strategy

Proposition

For every $\rho \in (0, \varphi)$, and for every value $\gamma > 0$, a unique equilibrium strategy $p_e \in [0, 1]$ exists.

This result arises in spite of the fact that the function $C_S(p)$ is not necessarily monotone:

Equilibrium Strategy

Proposition

For every $\rho \in (0, \varphi)$, and for every value $\gamma > 0$, a unique equilibrium strategy $p_e \in [0, 1]$ exists.

This result arises in spite of the fact that the function $C_S(p)$ is not necessarily monotone:

Equilibrium Strategy

Proposition

For every $\rho \in (0, \varphi)$, and for every value $\gamma > 0$, a unique equilibrium strategy $p_e \in [0, 1]$ exists.

This result arises in spite of the fact that the function $C_S(p)$ is not necessarily monotone:

Equilibrium Strategy

Proposition

For every $\rho \in (0, \varphi)$, and for every value $\gamma > 0$, a unique equilibrium strategy $p_e \in [0, 1]$ exists.

This result arises in spite of the fact that the function $C_S(p)$ is not necessarily monotone:

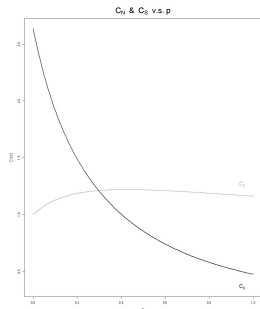


Figure: $\gamma = 1$ and $\rho = 0.725$

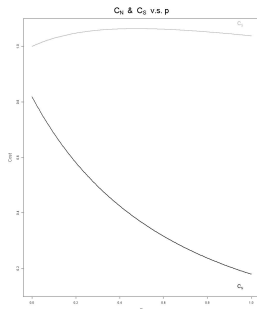


Figure: $\gamma = 1$ and $\rho = 0.45$

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

In equilibrium, it must hold that $C_S(p) = C_N(p)$, or equivalently:

$$\gamma \mathbb{E}[L(p) \mid Y = 0] = \frac{1}{\Pr(Y = 0)} = 1 + p\rho. \quad (8)$$

Proposition

The function $\mathbb{E}[L(p) \mid Y = 0]$ is continuous and monotone non-increasing in p .

To prove this we use the Sample Path Analysis technique, comparing two systems under the same sequence of events.

This property together with (8) immediately indicate that p_e is unique.

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim \mathcal{U}[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▷ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▷ $\tau_i \sim \exp(\mu)$;
 - ▷ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▶ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▶ $\tau_i \sim \exp(\mu)$;
 - ▶ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof.

Define:

- ▶ System $\Omega = \{S_Q, S_L, p\}$ with the state $(X(t), Y(t))$
- ▶ System $\Omega' = \{S'_Q, S'_L, p'\}$ with the state $(X'(t), Y'(t))$
- ▶ $\{(T_i, \tau_i, u_i)\}_{i \in \mathbb{N}}$ and $\forall i \in \mathbb{N}$:
 - ▷ $T_{i+1} - T_i \sim \exp(\Lambda)$;
 - ▷ $\tau_i \sim \exp(\mu)$;
 - ▷ $u_i \sim U[0, 1]$.

Assume, w.l.o.g that

- ▶ $p < p'$
- ▶ $(X(0), Y(0)) = (X'(0), Y'(0)) = (0, 0)$

We shall show:

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0] .$$

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

Equilibrium Strategy

Proof (Cont.)

We apply the following modifications:

- (i) If $Y(T_i) = 1$ (or $Y'(T_i) = 1$), customer i preempts the one in service, and the preempted customer is routed to S_Q (or S'_Q).
- (ii) Subsystem S_Q (or S'_Q) is a preemptive resume LCFS queue.

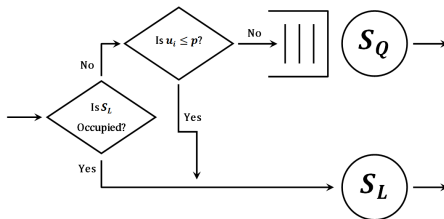


Figure: Customers' flow chart of the modified system

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customers i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Proof (Cont.)

- (a) If there is a customer in S_L (or S'_L), it must be the last customer.
- (b) Customer i begins service at T_i .
- (c) Joining $S_L \Rightarrow$ Joining S'_L

$$\forall t \in [0, \infty) : \{Y(t) = 1\} \Rightarrow \{Y'(t) = 1\};$$

- (d) Joining $S'_Q \Rightarrow$ Joining S_Q

$$\forall t \in [0, \infty) : \{Y'(t) = 0\} \Rightarrow \{Y(t) = 0\};$$

- (e) The sojourn time in $S'_Q \leq$ The sojourn time in S_Q

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$\mathbb{E}[X' \mid Y' = 0] \leq \mathbb{E}[X \mid Y' = 0],$$

and it is left to prove

$$\mathbb{E}[X \mid Y' = 0] \leq \mathbb{E}[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$\mathbb{E}_{Y=0}[X] = \lambda_1 \mathbb{E}[X \mid Y' = 0] + \lambda_2 \mathbb{E}_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Equilibrium Strategy

Proof (Cont.)

From (d) + (e),

$$\forall t \in [0, \infty) : X(t) \geq X'(t); \quad \text{or,} \quad X \succcurlyeq X'; \quad (9)$$

In fact,

$$E[X' \mid Y' = 0] \leq E[X \mid Y' = 0],$$

and it is left to prove

$$E[X \mid Y' = 0] \leq E[X \mid Y = 0].$$

Note that for some $\lambda_1, \lambda_2 \in [0, 1]$ s.t. $\lambda_1 + \lambda_2 = 1$:

$$E_{Y=0}[X] = \lambda_1 E[X \mid Y' = 0] + \lambda_2 E_{Y=0}[X \mid Y' = 1]. \quad (10)$$

Proof (Cont.)

In the paper we show explicitly that

$$E_{Y=0}[X] \leq E_{Y=0} \left[X \mid \begin{array}{l} \text{busy period} \\ \text{has begun} \end{array} \right] = E_{Y=0}[X \mid Y' = 1],$$

which, alongside (10) implies that

$$E[X \mid Y' = 0] \leq E_{Y=0}[X] \leq E_{Y=0}[X \mid Y' = 1],$$

to complete the proof of the proposition. \square

Proof (Cont.)

In the paper we show explicitly that

$$E_{Y=0}[X] \leq E_{Y=0} \left[X \mid \begin{array}{l} \text{busy period} \\ \text{has begun} \end{array} \right] = E_{Y=0}[X \mid Y' = 1],$$

which, alongside (10) implies that

$$E[X \mid Y' = 0] \leq E_{Y=0}[X] \leq E_{Y=0}[X \mid Y' = 1],$$

to complete the proof of the proposition. \square

Proof (Cont.)

In the paper we show explicitly that

$$E_{Y=0}[X] \leq E_{Y=0} \left[X \left| \begin{array}{l} \text{busy period} \\ \text{has begun} \end{array} \right. \right] = E_{Y=0}[X \mid Y' = 1] ,$$

which, alongside (10) implies that

$$E[X \mid Y' = 0] \leq E_{Y=0}[X] \leq E_{Y=0}[X \mid Y' = 1] ,$$

to complete the proof of the proposition. \square

Proof (Cont.)

In the paper we show explicitly that

$$E_{Y=0}[X] \leq E_{Y=0} \left[X \left| \begin{array}{l} \text{busy period} \\ \text{has begun} \end{array} \right. \right] = E_{Y=0}[X \mid Y' = 1] ,$$

which, alongside (10) implies that

$$E[X \mid Y' = 0] \leq E_{Y=0}[X] \leq E_{Y=0}[X \mid Y' = 1] ,$$

to complete the proof of the proposition. □

Proof (Cont.)

In the paper we show explicitly that

$$E_{Y=0}[X] \leq E_{Y=0} \left[X \left| \begin{array}{l} \text{busy period} \\ \text{has begun} \end{array} \right. \right] = E_{Y=0}[X \mid Y' = 1],$$

which, alongside (10) implies that

$$E[X \mid Y' = 0] \leq E_{Y=0}[X] \leq E_{Y=0}[X \mid Y' = 1],$$

to complete the proof of the proposition. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$E[L(0) \mid Y = 0] = E[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$E[L(0) \mid Y = 0] = E[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$\mathbb{E}[L(0) \mid Y = 0] = \mathbb{E}[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$\mathbb{E}[L(0) \mid Y = 0] = \mathbb{E}[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$E[L(0) \mid Y = 0] = E[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Proposition

The pure strategy $p = 0$ is an equilibrium strategy (in other words $p_e = 0$) iff:

$$\rho \leq \frac{1}{1 + \gamma} .$$

Proof.

This is immediate, as $p = 0$ is the M/M/1 regular case and

$$E[L(0) \mid Y = 0] = E[L(0)] = \frac{\rho}{1 - \rho} .$$

Substituting this in (8) and isolating ρ we get the desired result. □

Equilibrium Strategy

Equilibrium Strategy

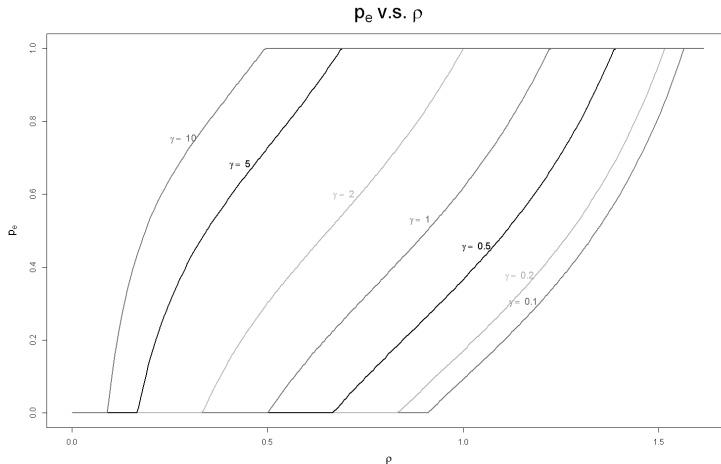


Figure: The equilibrium strategy p_e as a function of ρ for a various values of γ

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p) . \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p) . \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p) . \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p) . \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p). \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p). \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p). \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p). \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p). \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p). \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p). \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p). \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

The social objective function, $C(p)$, is defined as

$$C(p) := (1 - p)C_N(p) + pC_S(p). \quad (11)$$

Denote p^* the socially optimal strategy. Accordingly,

$$p^* := \arg \min_{p \in [0,1]} C(p). \quad (12)$$

and the values can be computed numerically.

For some values of ρ and γ , we spotted, counter intuitively that $p_e < p^*$.

Social Optimization

Strategic Sensing
in Cognitive Radio
Networks

Ran Snitkovsky

Social Optimization

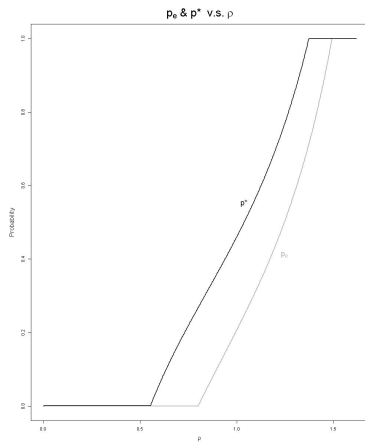


Figure: $\gamma = 0.25$

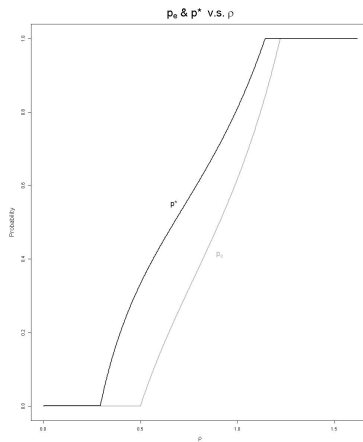


Figure: $\gamma = 1$

Social Optimization

Strategic Sensing
in Cognitive Radio
Networks

Ran Snitkovsky

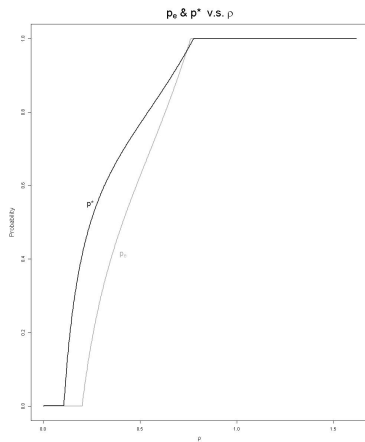


Figure: $\gamma = 4$

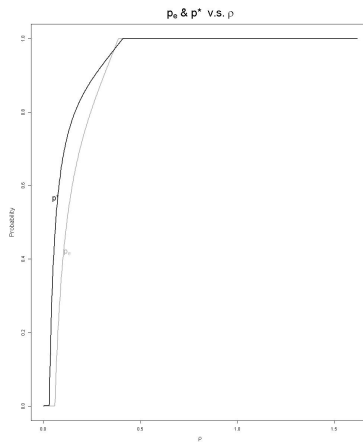


Figure: $\gamma = 16$

Social Optimization

Strategic Sensing
in Cognitive Radio
Networks

Ran Snitkovsky

Social Optimization

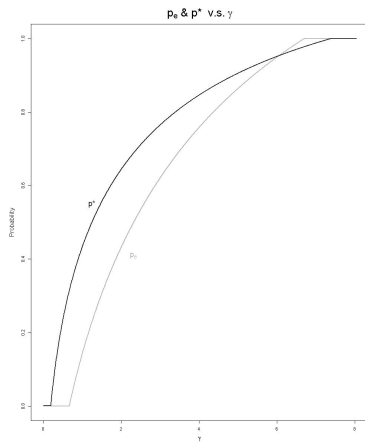


Figure: $\rho = 0.6$

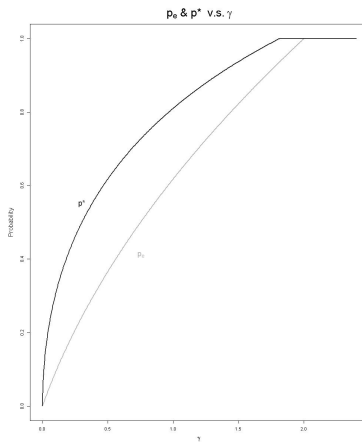


Figure: $\rho = 1$

Thank you for listening!