

Joint PHY-MAC Design for Opportunistic Spectrum Access with Multi-Channel Sensing

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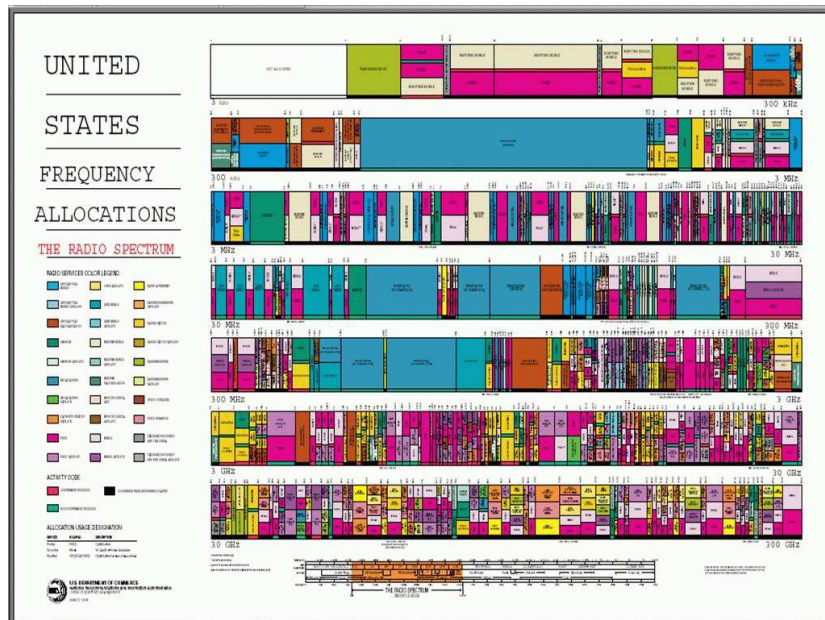
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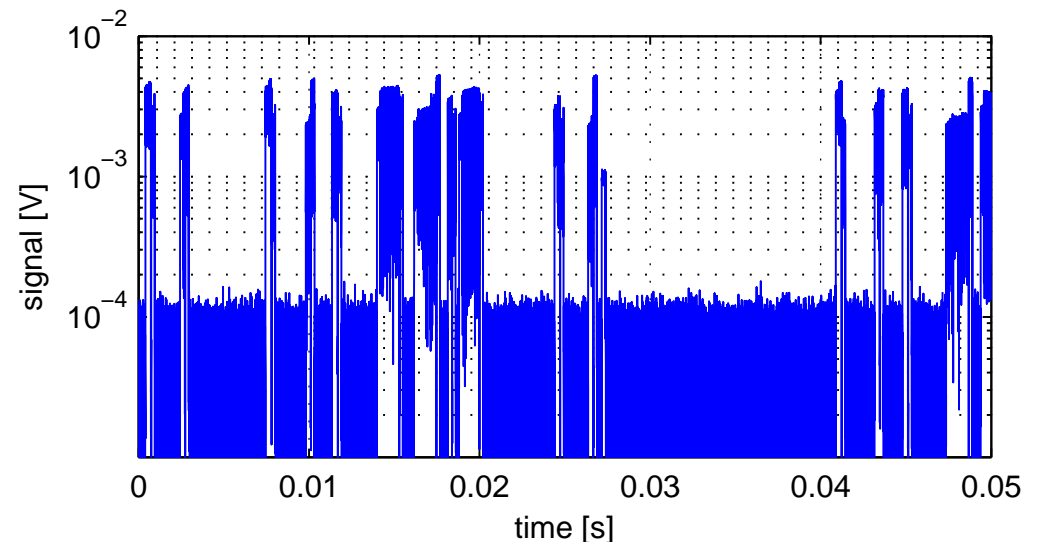
SPAWC'07 SESSION T2B: Cross-Layer Issues: From Physical to Networking Layers.

Spectrum Scarcity vs. Spectrum Opportunity

Overly Crowded Spectrum



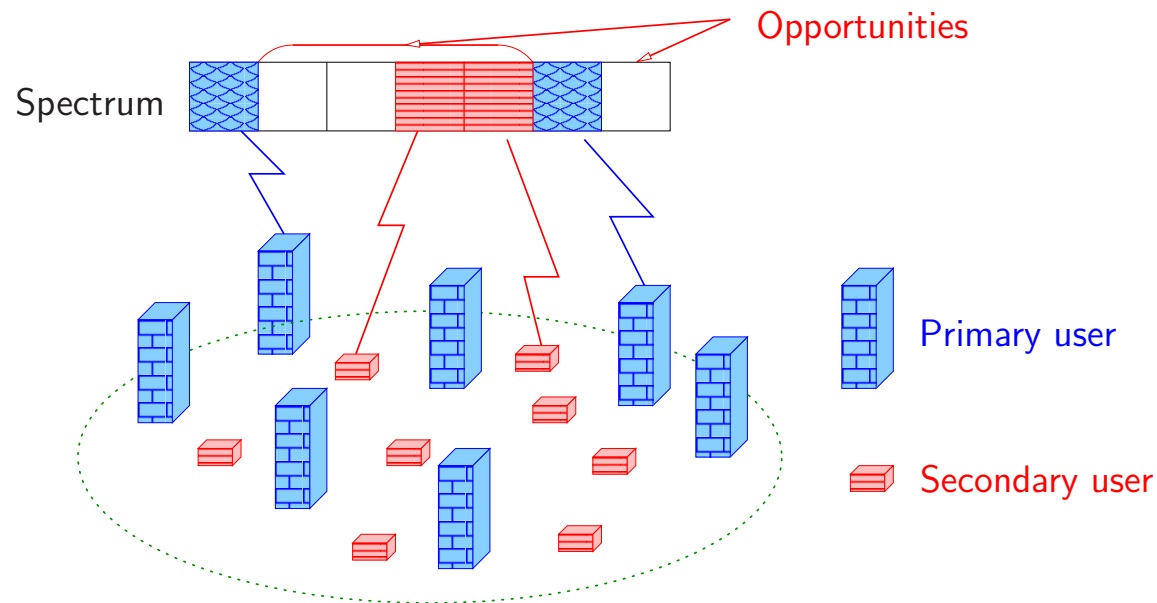
Pervasive Spectrum Opportunities



Spectrum usage of an active FTP session in a WLAN (ACSP at Cornell).

- ▶ Almost all usable radio frequencies have already been licensed.
- ▶ At any given time and location, a large portion of licensed spectrum lies unused.
 - Over 62% white space exists in the spectrum under 3GHz.
 - About 75% idle time during an active FTP session in WLAN.
 - Up to 90% idle time during voice-over-IP applications such as Skype.

Opportunistic Spectrum Access (OSA)



Basic idea:

Allow secondary users to exploit spectrum opportunities.

Design objective:

Maximize secondary users' throughput while limiting their interference to primary users (licensees).

Three basic design components:

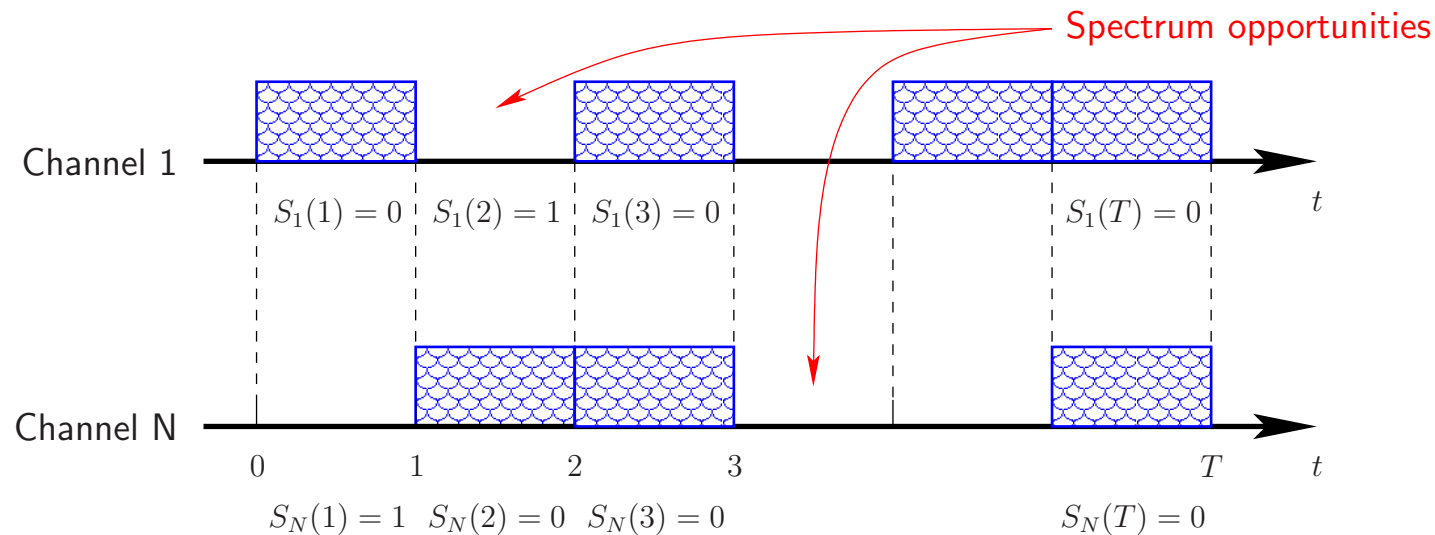
- ▶ **Spectrum sensor:** opportunity identification (PHY)
- ▶ **Sensing policy:** where in the spectrum to sense (MAC)
- ▶ **Access policy:** whether to tx given sensing errors may occur (MAC)

A decision-theoretic framework for joint PHY-MAC design.

Main Results

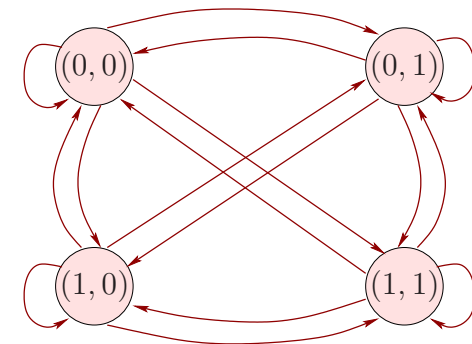
- ▶ A decision-theoretic framework based on partially observable Markov decision process (POMDP)
 - jointly optimizes the three basic design components
 - captures fundamental design tradeoffs
 - Spectrum sensor:** false alarm vs. miss detection
 - Access Policy:** overlooked opportunity vs. collision
 - Sensing Policy:** gaining access vs. gaining information
- ▶ Structural policies for the joint design
 - Separation principle for single-channel sensing and its extension to multi-channel sensing
 - Explicit optimal sensor design and closed-form optimal access policy.
 - Sensing design reduced from a constrained POMDP to an unconstrained one.
- ▶ Quantitative characterization of the interaction between PHY and MAC
 - Impact of the operating characteristics of spectrum sensor on MAC
 - Exploiting MAC information at PHY for improved sensor performance.

Network Model



- ▶ A spectrum of N channels, each with bandwidth B_n .
- ▶ A slotted primary network
 - Markovian spectrum usage with 2^N states:

$$\mathbf{S}(t) \triangleq [S_1(t), \dots, S_N(t)] \in \{0 \text{ (busy)}, 1 \text{ (idle)}\}^N.$$
 - Known transition probabilities.
- ▶ An ad hoc secondary network without dedicated control channel
 - Independent users, each can sense and access L channels in each slot.
 - User obtains a reward $R(t) = B_n$ for each successful access of idle channels.
 - User collides with primary users if access a busy channel.



Basic Components and Design Tradeoffs

Spectrum Sensor: false alarm vs. miss detection

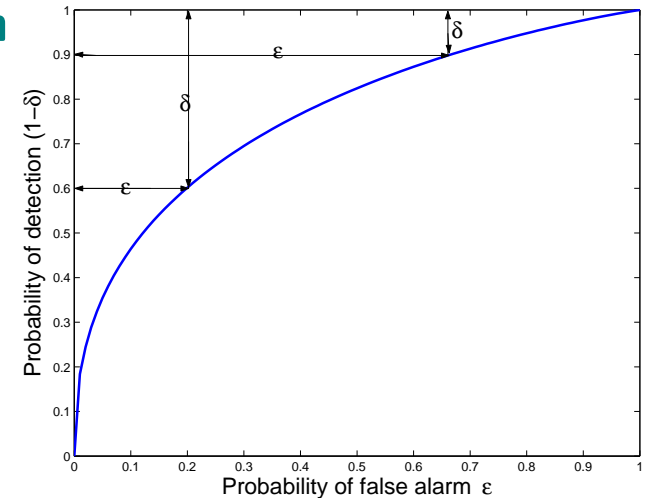
► Binary hypotheses test (for $L = 1$):

\mathcal{H}_0 : channel is idle vs. \mathcal{H}_1 : channel is busy

► Two Types of sensing errors:

□ false alarm (ϵ): $\mathcal{H}_0 \rightarrow \mathcal{H}_1$ (overlook)

□ miss detection (δ): $\mathcal{H}_1 \rightarrow \mathcal{H}_0$ (misidentification)



► Which point (ϵ, δ) on the ROC curve should the sensor operate? (decision rule)

Access Policy: overlooked opportunity vs. collision

► Consequences of trusting sensing outcome:

□ false alarm (idle sensed as busy) \Rightarrow overlooked opportunity

□ miss detection (busy sensed as idle) \Rightarrow collision

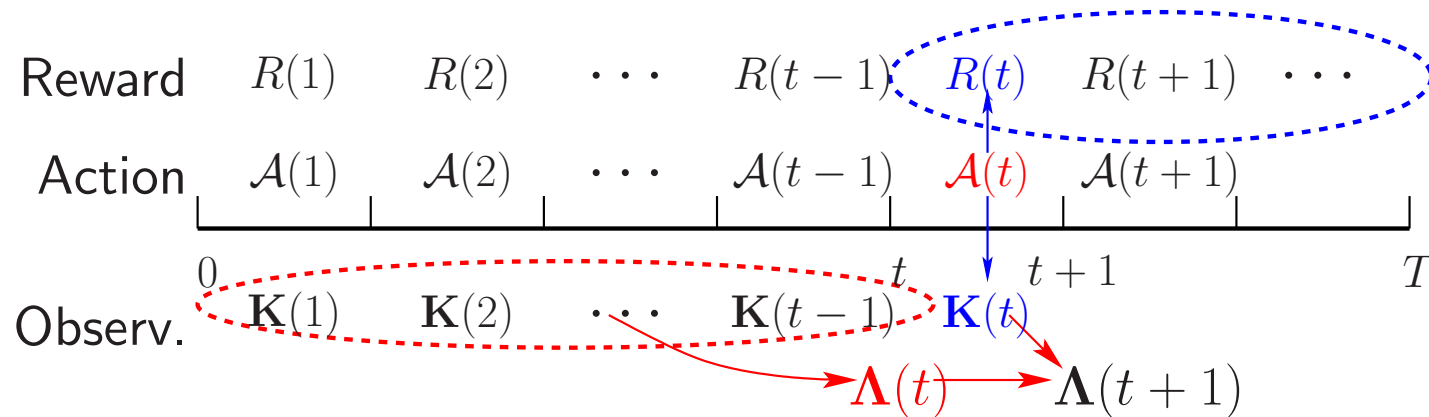
► When and how much to trust the sensor?

$$\text{For } L = 1, \text{ tx prob.} = \begin{cases} p_0 & \text{if idle} \\ p_1 & \text{if busy} \end{cases} \quad \begin{array}{l} p_0 < 1 : \text{conservative} \\ p_1 > 0 : \text{aggressive} \end{array} \Leftrightarrow \text{sensor } \{\epsilon, \delta\}$$

Joint design of access policy at MAC and spectrum sensor at PHY

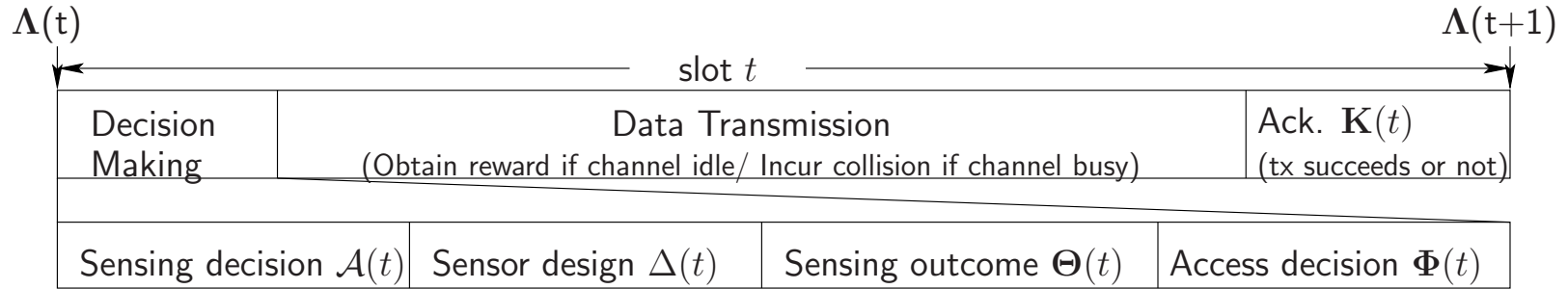
Basic Components and Design Tradeoffs

Sensing Policy: gaining access vs. gaining information



- ▶ Each observation $\mathbf{K}(t) \in \{0(\text{unsuccessful}), 1(\text{successful access})\}^L$ provides partial information on the spectrum usage state.
- ▶ Sensing action $\mathcal{A}(t)$ should be based on the **conditional distribution** $\Lambda(t)$ that exploits the entire decision and observation history.
- ▶ $\mathcal{A}(t)$ results in **immediate reward** $R(t)$ and **observation** $\mathbf{K}(t)$ that affects future reward.
- ▶ Optimal $\mathcal{A}(t)$ achieves the best tradeoff between gaining immediate reward and gaining spectrum information.

A Constrained POMDP Formulation



- ▶ **Belief vector** $\Lambda(t) = \{\Lambda_s(t)\}_{s \in \{0,1\}^N}$, where $\Lambda_s(t)$ is the conditional probability that the spectrum is in state s : $\{\Lambda(t), \mathcal{A}(t), \mathbf{K}(t)\} \rightarrow \Lambda(t+1)$
- ▶ **Sensing policy** π_s
 - deterministic: $\Lambda(t) \rightarrow$ a set $\mathcal{A}(t)$ of L channels to be sensed in slot t .
 - randomized: $\Lambda(t) \rightarrow$ PMF of $\mathcal{A}(t)$.
- ▶ **Spectrum sensor** π_δ : $\Lambda(t) \rightarrow$ a decision rule $\Delta(t)$ used for occupancy detection:

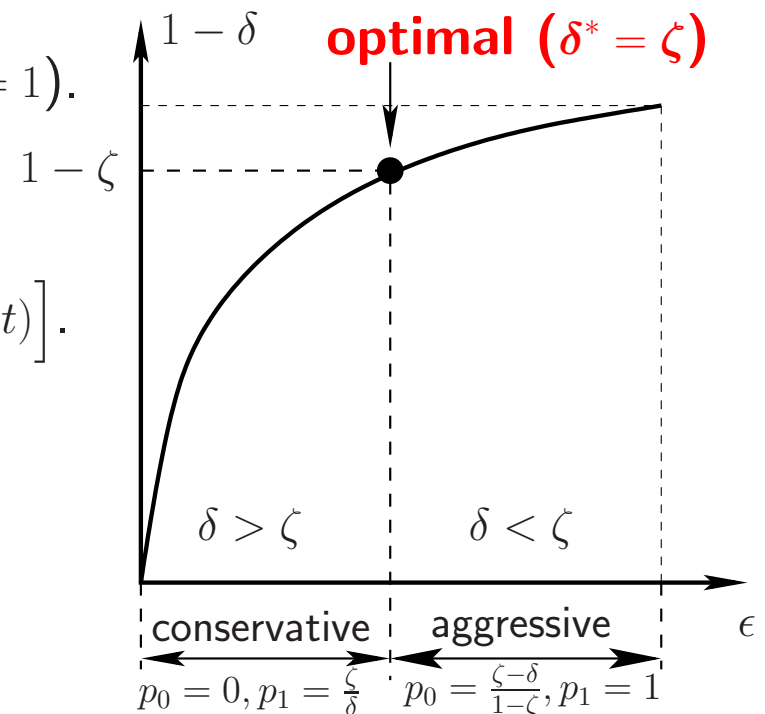
$$\{\Delta(t), \text{channel measurements}\} \rightarrow \Theta(t) = \{\Theta_n(t)\}_{n \in \mathcal{A}(t)} \in \{0(\text{busy}), 1(\text{idle})\}^L.$$
- ▶ **Access policy** π_c
 - deterministic: $\{\Lambda(t), \Theta(t)\} \rightarrow \Phi(t) \in \{0(\text{no access}), 1(\text{access})\}^L$.
 - randomized: $\{\Lambda(t), \Theta(t)\} \rightarrow$ tx probabilities.
- ▶ **Objective:**

$$\{\pi_\delta^*, \pi_s^*, \pi_c^*\} = \arg \max_{\pi_\delta, \pi_s, \pi_c} \mathbb{E} \left[\sum_{t=1}^T R(t) \right] \quad \text{s.t. collision prob. } P_n(t) \leq \zeta, \quad \forall n \in \mathcal{A}(t) \quad (*)$$

The Separation Principle for Single-Channel Sensing

Theorem: π_δ and π_c can be decoupled from π_s without losing optimality

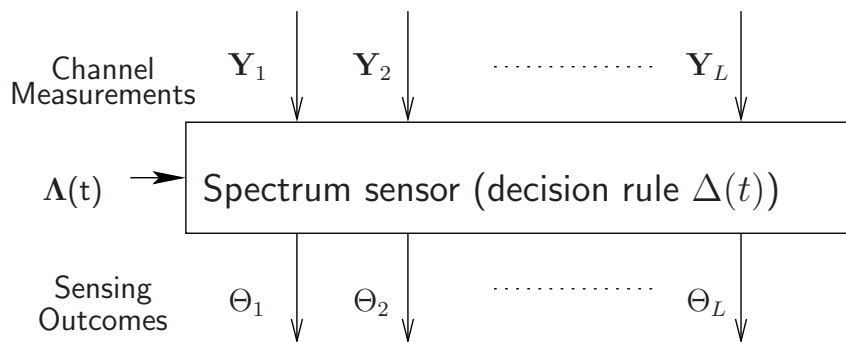
- Choose π_δ and π_c to max. immediate reward $R(t)$ and ensure constraint $P_n(t) = \zeta$.
 - ⇒ A static optimization problem.
 - ⇒ Explicit optimal design of spectrum sensor:
 - optimal Neyman-Pearson (NP) detector with prob. of missing (PM) $\delta = \zeta$.
 - ⇒ Closed-form optimal access policy:
 - trust sensing outcome (tx prob. $p_0 = 0, p_1 = 1$).
- Choose π_s to maximize total reward $\mathbb{E} \left[\sum_{t=1}^T R(t) \right]$.
 - ⇒ An unconstrained POMDP.
 - ⇒ Deterministic sensing policy.



Two Spectrum Sensor Structures for Multi-Channel Sensing

Goal: dynamically choose decision rules $\Delta(t)$ for spectrum opp. identification.

Joint opportunity identification



- ▶ Perform a 2^L -ary hypothesis test:

$$\mathcal{H}_0 : \mathbf{S}_{\mathcal{A}}(t) = [1, 1, \dots, 1],$$

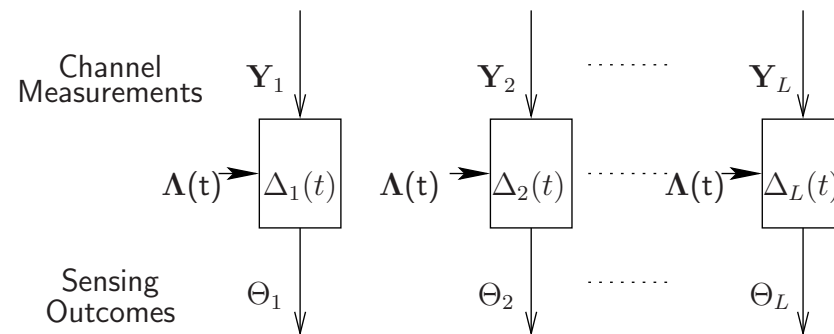
$$\mathcal{H}_1 : \mathbf{S}_{\mathcal{A}}(t) = [0, 1, \dots, 1],$$

...

$$\mathcal{H}_{2^L-1} : \mathbf{S}_{\mathcal{A}}(t) = [0, 0, \dots, 0].$$

- ▶ Decision rule: $\{\{\mathbf{Y}_n\}_{n=1}^L\} \rightarrow \{\mathcal{H}_0, \dots, \mathcal{H}_{2^L-1}\}$
jointly exploits channel measurements.
- ▶ Performance is specified by a set of $2^L \times (2^L - 1)$ error probabilities.

Independent opportunity identification

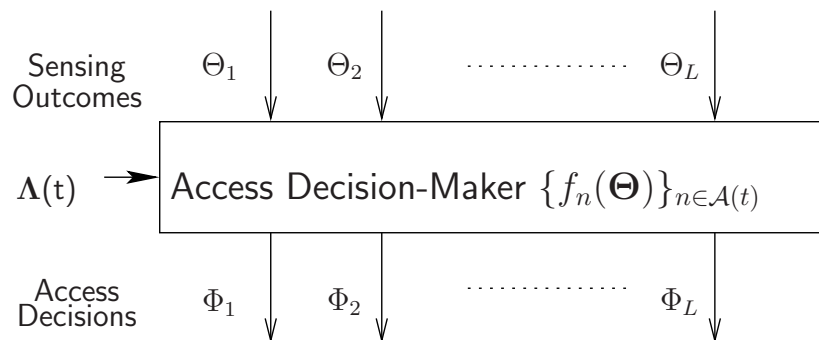


- ▶ Performs L independent hypothesis tests: $\mathcal{H}_0 : S_n(t) = 1,$
 $\mathcal{H}_1 : S_n(t) = 0, \quad n \in \mathcal{A}(t).$
- ▶ Decision rule $\Delta_n(t) : \{\mathbf{Y}_n\} \rightarrow \{\mathcal{H}_0, \mathcal{H}_1\}$
ignores correlation among channel measurements.
- ▶ Performance is specified by L pairs of false alarm and miss detection rates.
- ▶ Less complex than joint identification.

Two Access Policy Structures for Multi-Channel Sensing

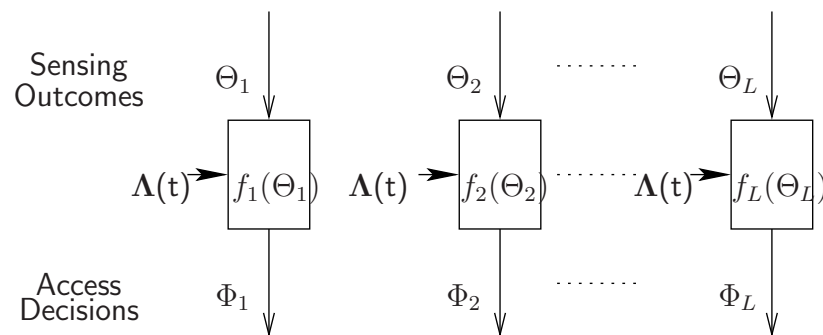
Goal: dynamically choose access decisions or transmission probabilities.

Joint access decision-making



- ▶ Tx. probability $f_n(\Theta) \triangleq \Pr\{\Phi_n = 1 | \Theta\}$ governs access decision $\Phi_n, \forall n \in \mathcal{A}(t)$.
- ▶ Access decision **jointly exploits sensing outcomes from all sensed channels:**
 $\Theta = \{\Theta_n\}_{n \in \mathcal{A}(t)}$.
- ▶ # of tx. probabilities to be designed = 2^L (possible sensing outcomes) $\times L$ (chosen channels).

Independent access decision-making



- ▶ Tx. probability $f_n(\Theta_n) \triangleq \Pr\{\Phi_n = 1 | \Theta_n\}$ independent of sensing outcomes from other channels.
- ▶ Access decision **ignores correlation among sensing outcomes.**
- ▶ # of tx. probabilities to be designed = $2L$ (chosen channels).
- ▶ Less complex than joint identification.

Extension of the Separation Principle

$$\{\pi_\delta^*, \pi_s^*, \pi_c^*\} = \arg \max_{\pi_\delta, \pi_s, \pi_c} \mathbb{E} \left[\sum_{t=1}^T R(t) \right] \quad \text{s.t. collision prob. } P_n(t) \leq \zeta, \quad \forall n \in \mathcal{A}(t) \quad (*)$$

Joint sensor & joint access structure

- ▶ Provides globally optimal solution.
- ▶ Requires **randomized** policies for optimality.
- ▶ Optimal but computationally prohibitive.

Independent sensor & independent access structure

- ▶ The **separation principle** holds.
- ▶ Optimal solution under this structure (the SP approach):
 - Spectrum sensor: optimal NP detector with PM = ζ .



- Access policy: trust the sensing outcome.

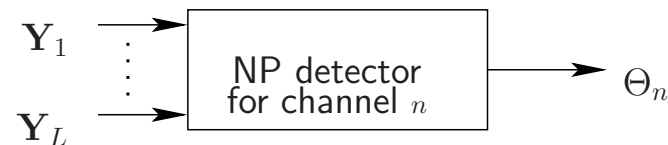


- ▶ Caveat: ignores correlation among channel occupancies.

Exploiting Correlation: The PHY Layer Approach

Joint sensor & independent access structure

- Sensor: optimal NP detector, using all channel measurements, with $PM = \zeta$.



- Access policy: trust the sensing outcome (using one sensing outcome).



- ▶ Exploits all channel measurements $\{\mathbf{Y}_n\}_{n \in \mathcal{A}(t)}$ in occupancy detection for each chosen channel.
- ▶ Uses MAC layer information at PHY layer: the *a priori* joint distribution of channel measurements is obtained from the **belief vector** $\Lambda(t)$.
- ▶ **Locally optimal** (maximizes instantaneous throughput).
- ▶ Reduces to the SP approach when channels evolve independently.

Exploiting Correlation: The MAC Layer Approach

Independent sensor & joint access structure

- Sensor: optimal NP detector, using single channel measurements, with $PM = \zeta$.



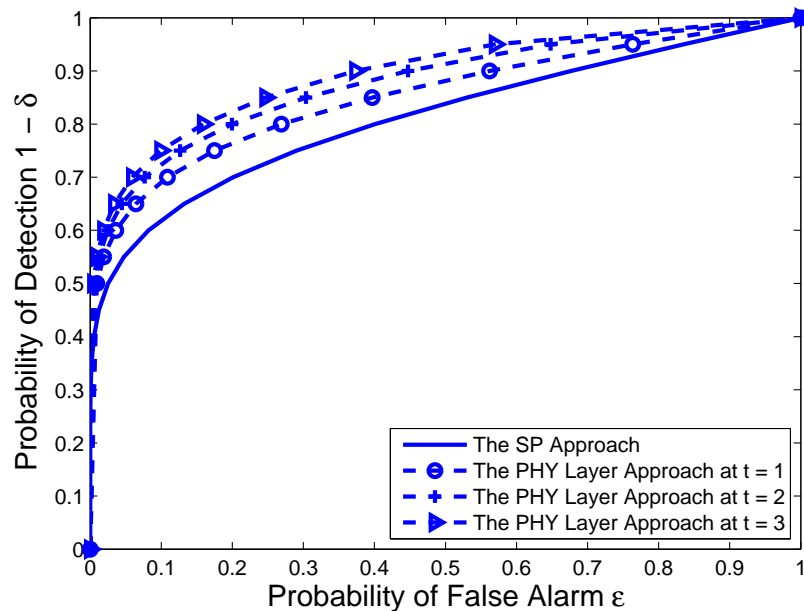
- Access policy: myopic tx probabilities $f_n(\Theta)$ (maximizes the instantaneous throughput) obtained via linear programming.



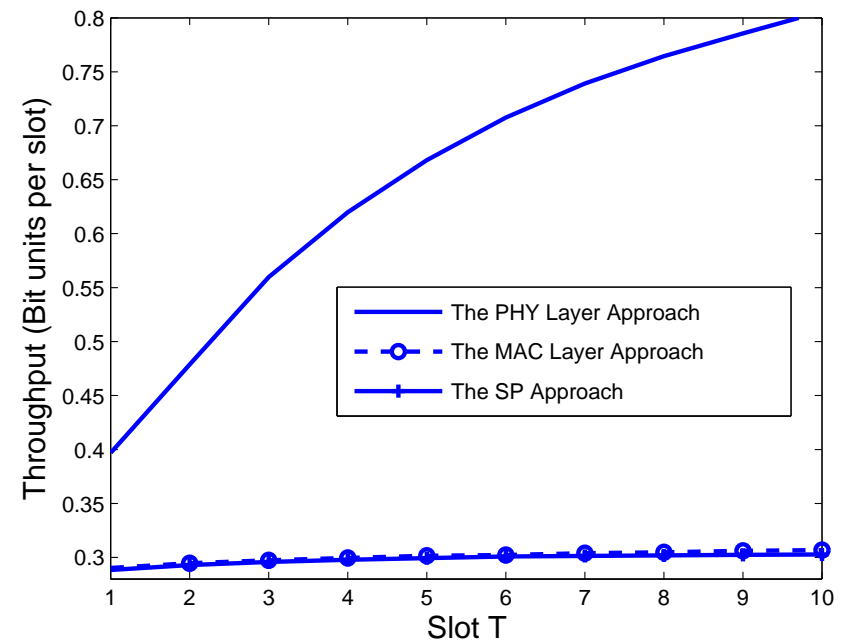
- ▶ Exploits all sensing outcomes $\Theta = \{\Theta_n\}_{n \in \mathcal{A}(t)}$ in making access decision for each chosen channel.
- ▶ **Locally optimal** when channels evolve independently.
- ▶ Reduces to the SP approach when channels evolve independently.

Performance Comparison

ROC of spectrum sensor



Normalized Throughput



- ▶ Performance of the PHY layer spectrum sensor is improved by exploiting the MAC layer sensing and access decisions (characterized by the belief vector).
- ▶ Exploitation of channel correlation at the PHY layer is more effective than that at the MAC layer.

Conclusions

- ▶ Optimal OSA with multi-channel sensing formulated as a constrained POMDP.
- ▶ **Separation Principle**
 - holds for $L = 1$; extends to $L > 1$ under the independent sensor & independent access structure.
 - leads to explicit optimal sensor design and closed-form optimal access policy.
 - reduces sensing design from a constrained POMDP to an unconstrained one.
- ▶ **Exploiting Channel Correlation**
 - PHY layer approach: improved sensor performance by exploiting MAC information (belief vector).
 - MAC layer approach: infers channel correlation from sensing outcomes.

Limitations

- ▶ Known transition probabilities of the underlying Markov process (robustness to model mismatch can be found in [1]).
- ▶ Interaction among secondary users not taken into account (exploited in [2]).

[1] Y. Chen, Q. Zhao, and A. Swami, "Joint Design and Separation Principle for Opportunistic Spectrum Access," in *Proc. of IEEE Asilomar Conference*, Nov. 2006.

[2] Y. Chen, Q. Zhao, and K. Liu, "Distributed Spectrum Sharing Among Competing Secondary Users," submitted to 45th Annual Allerton Conference on Communication, Control, and Computing, 2007.