





#### **A Multi-Agent Q-Learning Based Rendezvous Strategy for Cognitive Radios**

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#### 100 YEARS OF U.S. AIR FORCE SCIENCE & TECHNOLOGY

**Clifton Watson Air Force Research Laboratory**

*Integrity Service Excellence*







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- **Public and private sectors rely on spectrum access.**
- **Increasing demands require efficient spectrum use.**
- **This can be provided by cognitive radios (CRs) that can**
	- **sense, learn, and adapt to spectrum**
	- **access unused/underused licensed spectrum as unlicensed secondary users (SUs)**







- **SUs must quickly find each other to communicate in multi-channel spectrum environment.** 
	- **No dedicated control channel or central entity can be used.**
	- **All channels can be used for rendezvous and data exchange.**
- **Challenges**
	- **PU and SU activities are random and unpredictable.**
		- **Minimize PU interference (PUI)**
		- **Avoid SU collisions**







- **Channel hopping (CH) is most common approach.**
	- **Predetermined CH sequences**
		- **Not biased towards any channels**
		- **Vulnerable to PUI and collisions**
	- **Adaptive CH sequences**
		- **Biased towards channels with least detected PU activity**
		- **Robust to PUI but more vulnerable to SU collisions**









- **Multi-Agent Q-Learning Rendezvous (MAQLR) Strategy**
	- **SUs actively learn which channels are best for rendezvous.**
	- **Learning is based on exploration of spectrum environment.**
	- **Learned channels are generally less prone to PUI and SU collisions.**





## **System Model & Assumptions**

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**Primary User** 

 $(\cdot_1)$ 

Channel 1 Channel 2

Channel M

 $((\cdot))$ 

Secondary User

Secondary Link

 $\left(\left(\cdot\right)\right)$ 



- N SU pairs (sender and receiver SU)
- M licensed channels ( $1 \le m \le M$ )
- Localized channel availability  $(\theta_m)$
- Slotted channel access by PUs and SUs
- Rendezvous in single slot with RTS/CTS
	- SU Assumptions
		- start rendezvous at same time
		- sense the same PU activity
		- can distinguish between PU and SU
		- can access one channel at a time
		- do not exchange info with each other







### **Application of Q-Learning to Blind Rendezvous (Cont'd)**



- **Reward Strategy**
	- **Reward received for channel at end of slot** 
		- $\int$  0, PU activity is detected
	- $\hskip.08cm -\quad R_{m}(j) = \left\{ \, r \right.$  PU activity is not detected and rendezvous fails
		- , PU activity is not detected and rendezvous succeeds
	- $-$  *r* **is a random reward,**  $(0 < r < 1)$ **<sup>\*</sup>**
- **Rendezvous fails when**
	- **RTS not transmitted**
	- **Paired SUs select different channels**
	- **Collision occurs**
	- **Poor channel quality**
- **Rewards capture dynamics of PU and SU activity, as well as channel quality.**
- **\* For sake of brevity, audience is referred to paper for computation of .**





### **Application of Q-Learning to Blind Rendezvous (Cont'd)**



- Q-values updated for channel m at end of slot *i* 
	- $Q_m(j + 1) = (1 \alpha_m(j))Q_m(j) + \alpha_m(j)R_m(j)$
	- $\alpha_m(i)$  is the learning rate,  $(0 \le \alpha_m(i) \le 1)$ 
		- **determines how much old info is valued over new info**
		- **starts at 1 and decreases over time**
- Probability of selecting channel m in slot *i*

$$
- P_m(j) = \frac{e^{Q_m(j)}/w}{\sum_m e^{Q_m(j)}/w}
$$

- **balances tradeoff between exploration and exploitation**
- **is temperature parameter**
	- **decreases exploration over time**
	- $\cdot$  set to  $\alpha_{m}(i)$  to decrease exploration by learning rate











## **Simulation Results**



- **Simulation Setup**
	- **5-channel spectrum environment with localized availability**
	- $-\theta_1 = 0.5, \theta_2 = 0.3, \theta_3 = 0.6, \theta_4 = 0.4, \theta_5 = 0.7$
	- $s = 0.9$
	- $-$  p varies: 0.7, 0.8, 0.9, 1
	- **Number of SU pairs vary from 1 to 10.**
	- **Sender SUs assumed to always have data to send.**
	- **Compared against existing adaptive techniques**
		- **Enhanced Adaptive Multiple Rendezvous Control Channel with Variable Slots (EAMRCC-VS)**
		- **Nested Grid Quorum Frequency Hopping (NGQFH)**
		- **Follow same sensing procedure as MAQLR strategy**







• Numbers of SU pairs on learned channels  $(s = 0.9, p = 0.9)$ 



Distribution A. Approved for public release; distribution unlimited . 88ABW-2017-1367 • SUs learn to use channels in effective and efficient manner.







- **SU Collision Rate (SUCR)**
	- **Average number of SU collisions per RTS transmission**



• SU's use of channels cause MAQLR strategy to have lower SUCR.



# **Simulation Results (Cont'd)**



- **PUI Rate**
	- **Average number of PUIs per RTS transmission**



• SU's use of channels result in slightly higher PUIR for MAQLR strategy.







- **Average Time-to-Rendezvous (ATTR)**
	- **Average number of slots to complete RTS/CTS handshake**



• MAQLR strategy has much lower ATTR mainly because of lower SUCR.





- **Throughput Efficiency (TE)**
	- **Ratio of actual throughput and maximum achievable throughput**
	- **Throughput is DATA packets exchanged per time slot.**



• MAQLR strategy has higher TE primarily because of lower SUCR.





- **SUs enhance rendezvous performance with MAQLR strategy.**
	- **Actively learns which channels are best for rendezvous**
	- **Learns channels based on perceived PU activities and rendezvous successes/failures**
	- **Learns how to access channels effectively and efficiently**
- **Enhanced performance comes at the cost of higher PUIR.**
- **Future plans to improve strategy by lowering PUIR while still achieving desired (if not better) performance.**





### **Questions**





