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

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#### **Outline**



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#### Introduction



- 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)





### **Blind Rendezvous Problem**



- 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





## **Existing Approaches**



- 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





## **Proposed Method: A MAQLR Strategy**



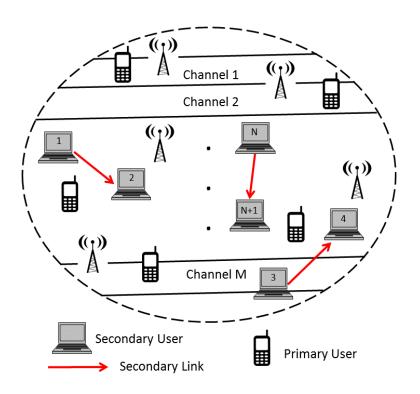
- Multi-Agent Q-Learning Rendezvous (MAQLR)
  Strategy
  - SUs actively learn which channels are best for rendezvous.
  - Learning is based on exploration of spectrum environment.
  - Learned channels are generally less prone to PUI and SU collisions.





## **System Model & Assumptions**





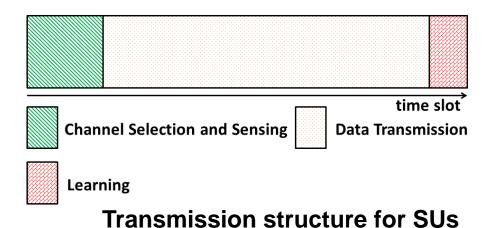
- N SU pairs (sender and receiver SU)
- M licensed channels  $(1 \le m \le M)$
- Localized channel availability  $(\theta_m)$
- Slotted channel access by PUs and SUs
- Rendezvous in single slot with RTS/CTS
- SU Assumptions
  - start rendezvous at same time
  - sense the same PU activity
  - can distinguish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



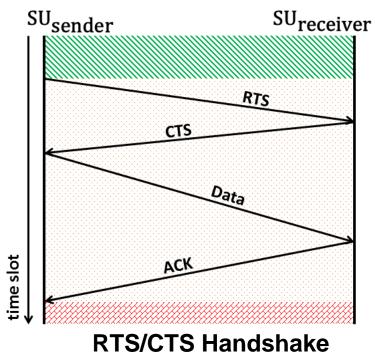


## System Model & Assumptions (cont'd)





- SUs sense correctly with probability s.
- Sender SUs transmit RTS with probability p.







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



- Reward Strategy
  - Reward received for channel m at end of slot j

$$- R_m(j) = \begin{cases} 0, \text{ PU activity is detected} \\ r, \text{ PU activity is not detected and rendezvous fails} \\ 1, \text{ PU activity is not detected and rendezvous succeeds} \end{cases}$$

- r is a random reward, (0 < r < 1) \*
- 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 r.





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



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

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

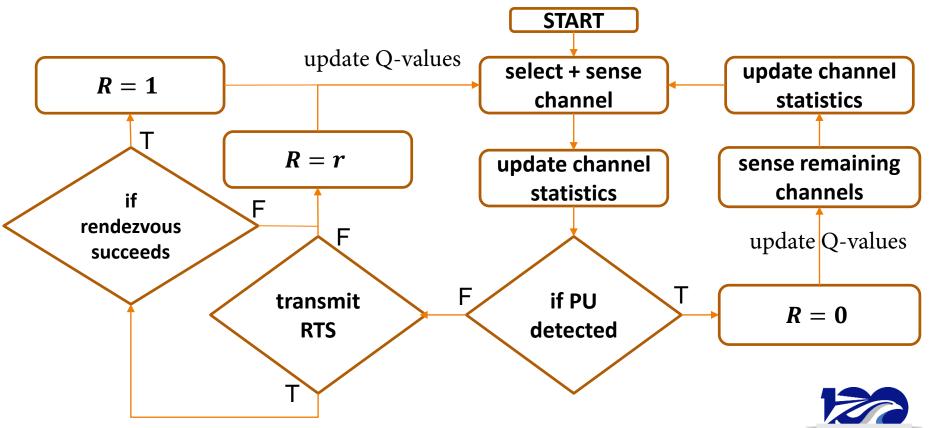
- balances tradeoff between exploration and exploitation
- W is temperature parameter
  - · decreases exploration over time
  - set to  $\alpha_m(j)$  to decrease exploration by learning rate





## **MAQLR Strategy - Description**







#### **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)

No.	$\theta_1 = 0.5$	$\theta_2 = 0.3$	$\theta_3 = 0.6$	$\theta_4 = 0.4$	$\theta_5 = 0.7$
1	0	0	0	0	1
2	0	0	1	0	1
3	1	0	1	0	1
4	1	0	1	1	1
7	1	1	2	1	2
8	2	1	2	1	2
9	2	1	2	2	2

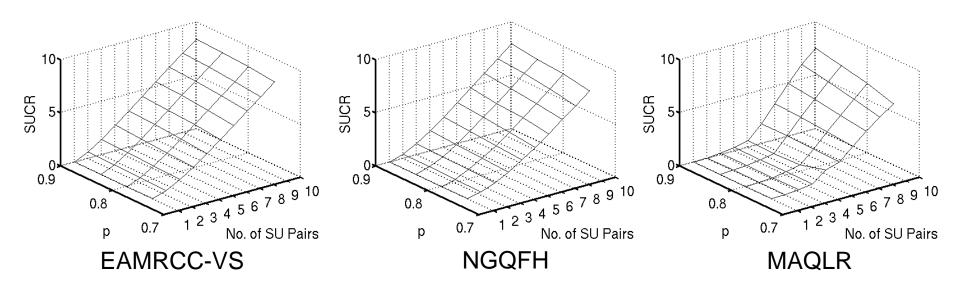
SUs learn to use channels in effective and efficient manner.







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



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

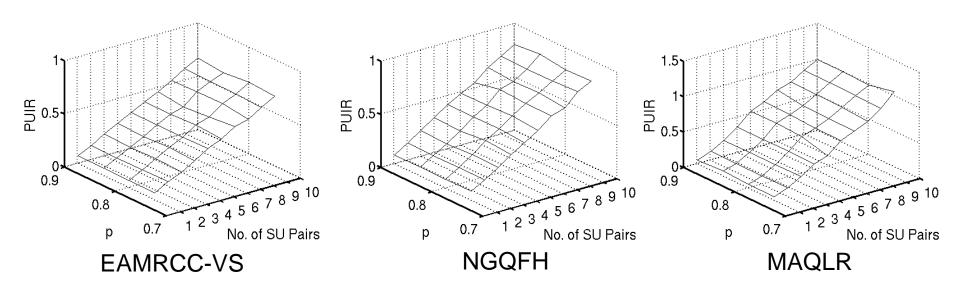






#### PUI Rate

Average number of PUIs per RTS transmission



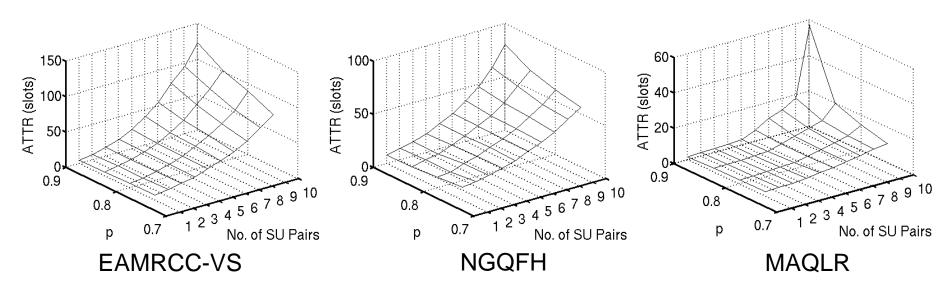
SU's use of channels result in slightly higher PUIR for MAQLR strategy.







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



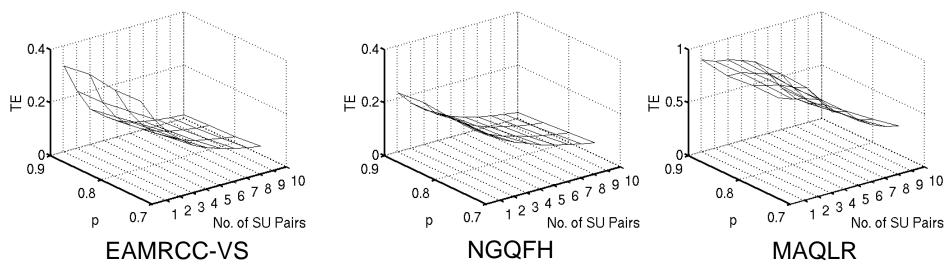
MAQLR strategy has much lower ATTR mainly because of lower SUCR.







- 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.





### Conclusion



- 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**



