Multi-Antenna Interference Cancellation Techniques for Cognitive Radio Applications

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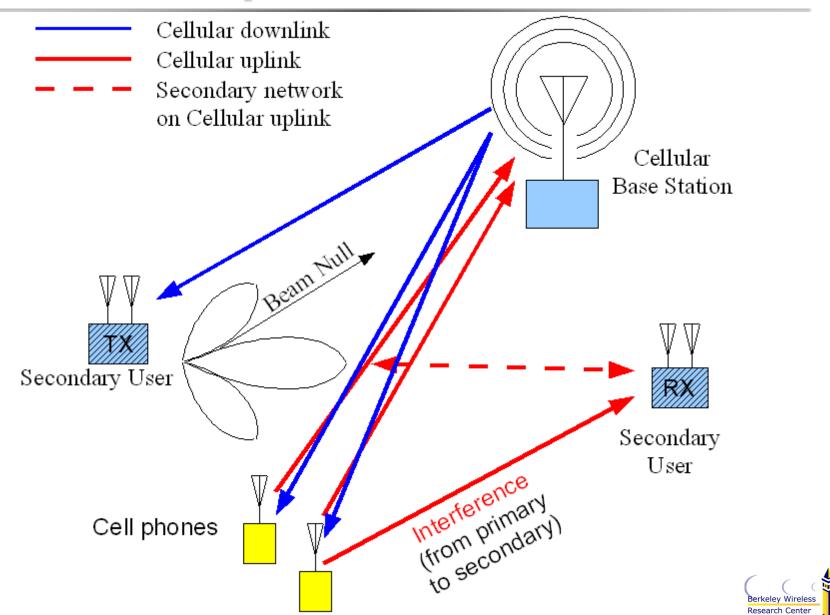
Last Time

- Improving spectrum reuse using primary and secondary collaboration¹
- More effective spatial reuse using multiple antennas on the secondary
- Example: cellular uplink reuse
- Today: signal processing and array processing techniques to improve collaboration
- To appear in IEEE WCNC 2009

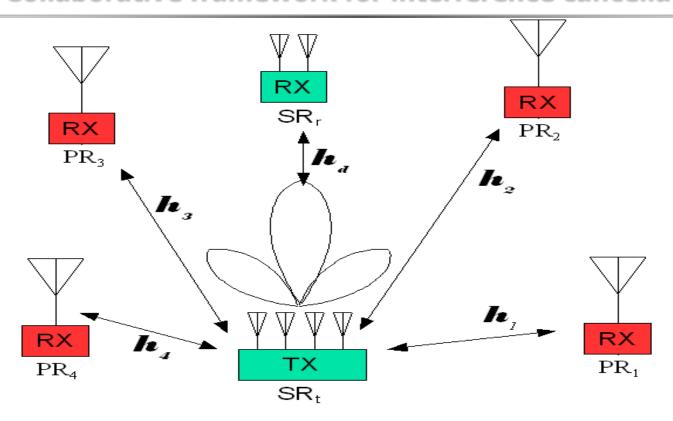
¹O. Bakr, M. Johnson, B. Wild, and K. Ramchandran, "A multi-antenna framework for spectrum reuse based on primary-secondary cooperation," in *IEEE Symposium on New Frontiers in Dynamic Spectrum Access Networks (DySPAN)*, October 2008.



Cellular uplink reuse framework



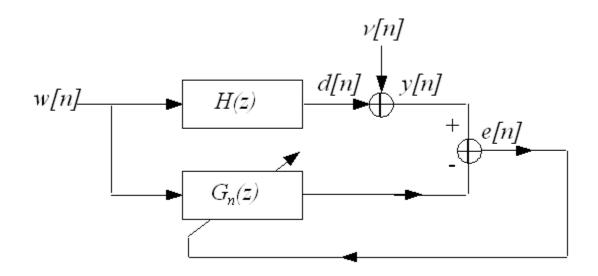
Collaborative framework for interference cancellation



- • h_j for 0 < j < K+1 are the channel responses from the secondary transmitter (SR_t) to each of K primary users (base stations) respectively.
- • h_d is the channel response from SR_t to SR_r .
- •Choose c to be the component of h_d that is orthogonal to h_j for 0 < j < K+1 (projection)
- •Channels unknown apriori? Need to estimate.

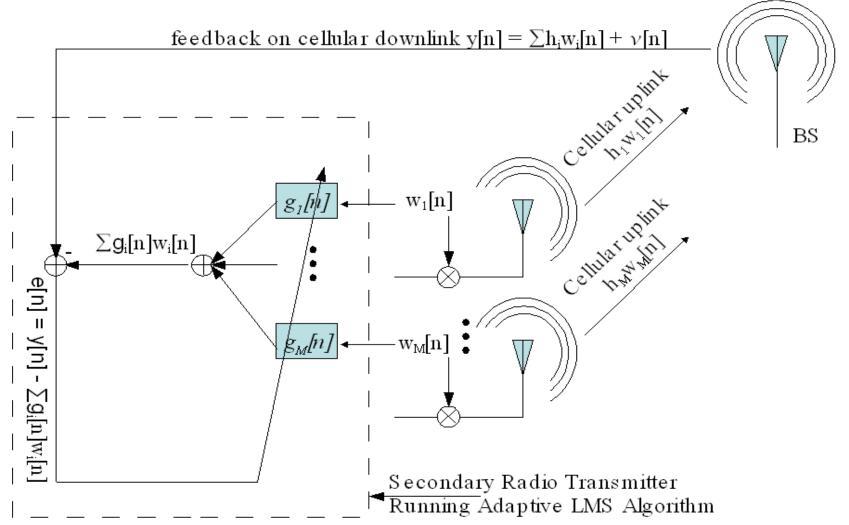


Estimation using adaptive filtering



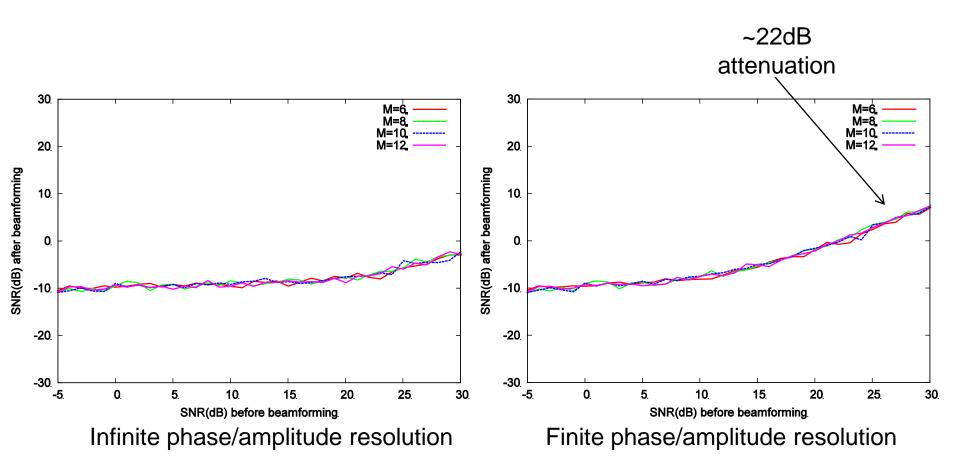
- •Identifying an unknown filter (channel) H(z) using an adaptive filter (e.g. Least Mean Square (LMS) algorithm)
- •w[n] is a known pseudo random sequence, $G_n(z)$ is the local estimate
- • $G_n(z)$ will converge to a noisy estimate of H(z) (due to the presence of noise)
- •In the beamforming context, the taps of H(z) are the complex responses from each antenna element on the secondary radio towards a primary radio

Beam-nulling using adaptive filtering





Simulated interference rejection



4 primary users (base stations), cognitive radio has M antennas

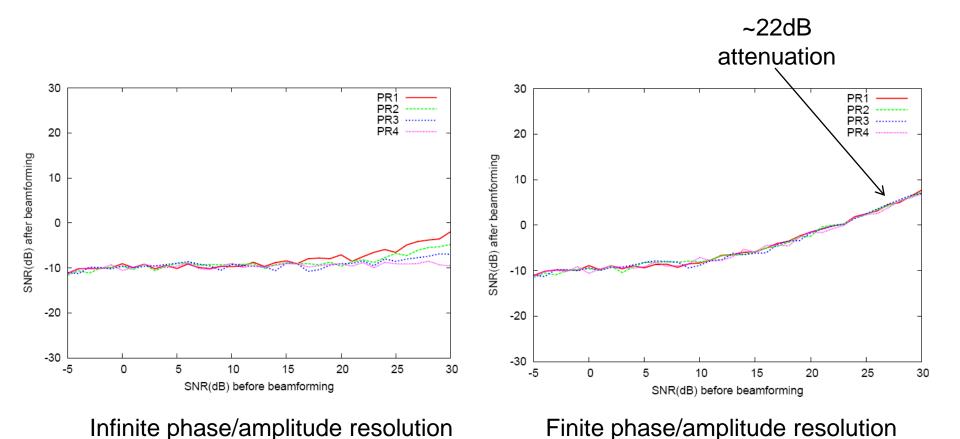


Iterative channel estimation

- Less coordination among primary users
- Better reuse of allocated channels
- Same adaptive algorithm, different choice of training sequence w
- Adaptively perform a Gram-Schmidt orthogonalization
 - Start with the closest node (e.g. PR1)
 - Run LMS at low power (no interference to other nodes)
 - After estimating h_1 , increase the power and choose w orthogonal to h_1
 - This will estimate the component of h_2 orthogonal to h_1
 - Increase the power, choose w orthogonal to both h_1 , h_2



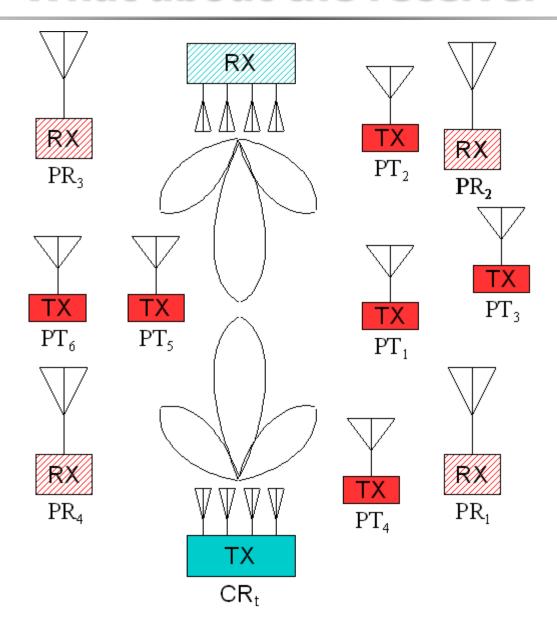
Simulated interference rejection



4 primary users (base stations), cognitive radio has 12 antennas



What about the receiver





Interference suppression framework

- Primary transmitters (cell phones) can cause interference to the secondary (cognitive radio) network
- At the secondary receiver:
 - $y[n] = \boldsymbol{h}_{d}\boldsymbol{d}[n] + \sum_{i} \boldsymbol{h}_{i}\boldsymbol{d}_{i}[n] + \boldsymbol{v}[n]$
- Choose beamforming (spatial filter) to maximize SINR
 - SINR_{out}= $|\boldsymbol{c}^h \boldsymbol{h}_d|^2/(\sum_i |\boldsymbol{c}^h \boldsymbol{h}_d|^2 + N_v)$
- MMSE criterion:
 - $c_{MMSE} = \text{arg min}_c |e[n]|^2 = \text{arg min}_c |c^h y[n] d[n]|^2$
- Good rejection in slow fading channels
- Doppler/frequency offsets can create problems

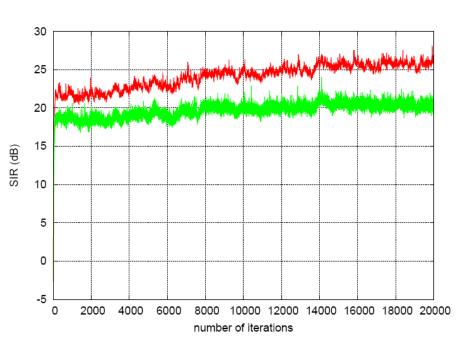


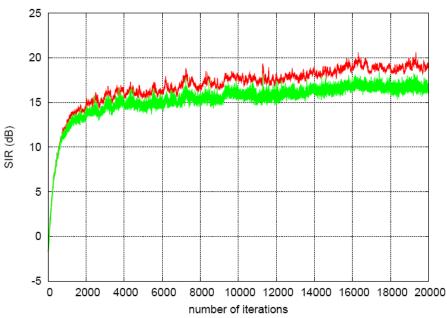
Differential MMSE Framework²

- Even in fast fading environments, channel remains relatively constant over successive symbols
- Avoid tracking channel variations by only looking at the difference (ratio) between symbols (similar to differential modulation)
- DMMSE criterion:
 - c_{DMMSE} = arg min_c | $c^h y[n-1]d[n] c^h y[n]d[n-1] |^2$
 - Subject to $E[|c^hy[n]|^2]=1$

²U. Madhow, K. Bruvold, and L. J. Zhu, "Differential MMSE: A framework for robust adaptive interference suppression for ds-cdma over fading channels," *IEEE Transactions on Communications, vol. 53, no. 8,* pp. 1377–1390, Aug. 2005.

Simulated interference rejection





DMMSE, offset=10% symbol rate

NLMS, offset=1% symbol rate

8 antennas, 6 interferers, SNR_{in}=-20dB, SIR_{in}=-40dB

