

Main Presentation

User Activity Detection in PDMA

Chirag Ramesh
SPC Lab, Indian Institute of Science

May 04, 2019

Contents

1. Recap

1.1 PDMA

1.2 System Model

1.3 Channel Estimation

2. Results

3. User Activity Detection

3.1 ML/MAP

3.2 CS methods

3.3 Channel Hardening & CS

4. References

What is PDMA?

- Pattern Division Multiple Access
- Replication of packets across resource elements
- Enabled via **successive interference cancellation**
- With capture effect, throughput can be increased to greater than 1
- PDMA - binary patterns assigned to users

	User 1	User 2	User 3	User 4	User 5	User 6
Slot 1	✓	✓		✓		
Slot 2	✓		✓		✓	
Slot 3		✓				✓
Slot 4			✓			

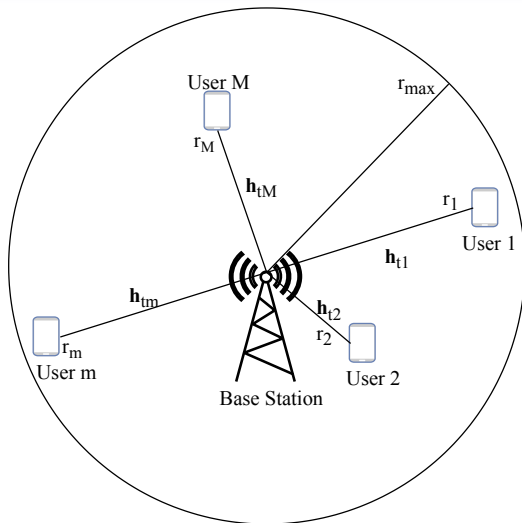
$$\mathbf{G} = \begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1M} \\ g_{21} & g_{22} & \cdots & g_{2M} \\ \vdots & \vdots & \ddots & \vdots \\ g_{T1} & g_{T2} & \cdots & g_{TM} \end{bmatrix}$$

System Model

- Users transmit packet replicas according to the pattern matrix \mathbf{G}
- $\mathbf{y}_t = \sum_{m=1}^M g_{tm} \mathbf{h}_{tm} x_m + \mathbf{n}_t$
- Noise, $\mathbf{n}_t \sim \mathcal{CN}(\mathbf{0}_N, N_0 \mathbf{I}_N)$
- Transmit packets x_m have $\mathbb{E}[x_m] = 0$ & $\mathbb{E}[|x_m|^2] = P$
- Channel gain,

$$\mathbf{h}_{tm} = \sqrt{\beta_m} \mathbf{v}_{tm}$$
- Path loss, $\beta_m = (r_m/r_0)^{-\alpha}$
- Fading,

$$\mathbf{v}_{tm} \sim \mathcal{CN}(\mathbf{0}_N, \sigma_h^2 \mathbf{I}_N)$$



Channel Estimation

- Each packet header carries a pilot $x_m = \sqrt{P}$

$$\mathbf{y}_t^{pk} = \sum_{i \in \mathcal{S}_k} g_{ti} \mathbf{h}_{ti} \sqrt{P} + \mathbf{n}_t^p \quad (1)$$

- MMSE channel estimates are recomputed every iteration as

$$\hat{\mathbf{h}}_{tm}^k \triangleq \mathbb{E}_{\mathbf{y}_t^{pk}} [\mathbf{h}_{tm}]$$

$$\hat{\mathbf{h}}_{tm}^k = \frac{\sqrt{P} \sigma_h^2 g_{tm} \beta_m}{P \sigma_h^2 (\sum_{i \in \mathcal{S}_k} g_{ti}^2 \beta_i) + N_0} \mathbf{y}_t^{pk} =: \eta_{tm}^k \mathbf{y}_t^{pk} \quad (2)$$

- The estimation error $\tilde{\mathbf{h}}_{tm}^k \triangleq \hat{\mathbf{h}}_{tm}^k - \mathbf{h}_{tm}$ is uncorrelated with the received pilot signal and the estimate itself

- *Threshold based decoding*: $\text{SINR} \geq \gamma_{th} \geq 1$
- Under the common pilots scheme, MRC is performed

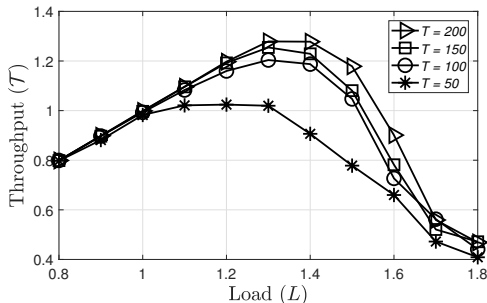
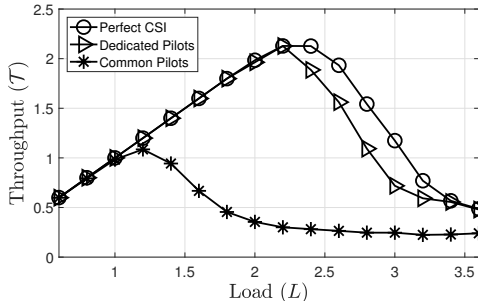
$$\begin{aligned} \tilde{y}_{tm}^k &= \hat{\mathbf{h}}_{tm}^{kH} \mathbf{y}_t^k = \underbrace{\hat{\mathbf{h}}_{tm}^{kH} \hat{\mathbf{h}}_{tm}^k g_{tm} x_m}_{\text{Signal}} + \underbrace{\hat{\mathbf{h}}_{tm}^{kH} \mathbf{n}_t}_{\text{Noise}} \\ &+ \underbrace{\hat{\mathbf{h}}_{tm}^{kH} \sum_{i \in \mathcal{S}_k^m} g_{ti} \mathbf{h}_{ti} x_i - \hat{\mathbf{h}}_{tm}^{kH} \tilde{\mathbf{h}}_{tm}^k g_{tm} x_m}_{\text{Interference}} \end{aligned} \quad (3)$$

- Effective SINR is calculated as

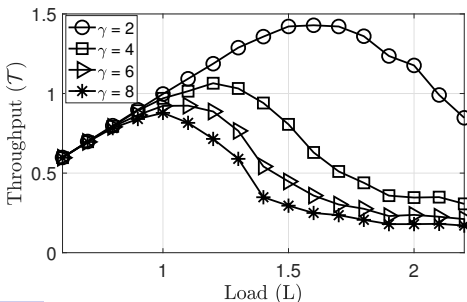
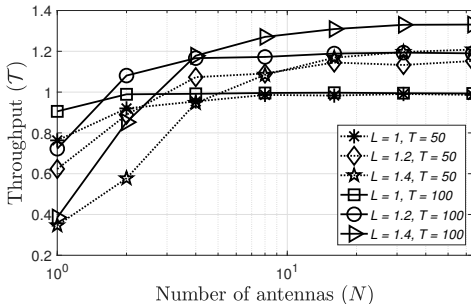
$$\text{SINR}_{tm}^k = \frac{P g_{tm}^2 \|\hat{\mathbf{h}}_{tm}^k\|^2}{N_0 + P \left(\sum_{i \in \mathcal{S}_k} g_{ti}^2 \delta_{ti}^k + \sum_{i \in \mathcal{S}_k^m} g_{ti}^2 \|\hat{\mathbf{h}}_{ti}^k\|^2 \right)} \quad (4)$$

Results

- Parameters: $P = 1$, $\gamma = 4$,
 $\alpha = 3$, $N = 4$,
 $\sigma_h^2 = 1$, $T = 50$,
 $r_{max} = 1\text{km}$, $r_0 = 0.1\text{km}$,
 $N_0 = 10^{-4}$ & $N_{sim} = 100$
- Common pilots result in **drop of 50% throughput** w.r.t. of perfect CSI
- Dedicated pilots can recover loss at the cost of $M\bar{d} - T = 190$ **extra training symbols** for $M = 60$



- For lower loads, increasing the number of antennas marginally increases the throughput and saturates thereafter
- For higher loads, increasing the number of antennas increases the throughput
- Decreasing the capture threshold yields higher throughputs as more users are decoded



User Activity Detection

Approach 1:

- Received Pilot: $\mathbf{y}_t^{pk} = \sqrt{P} \sum_{i \in \mathcal{S}_k} g_{ti} a_i \mathbf{h}_{ti} + \mathbf{n}_t^p$
- ML/MAP based detection

$$\mathbf{y}_t^{pk} \mid \mathbf{a} \sim \mathcal{CN}(\mathbf{0}_N, s_t \mathbf{I}_N)$$

$$s_t = P \sigma_h^2 \left(\sum_{i \in \mathcal{S}_k} g_{ti}^2 a_i^2 \beta_i \right) + N_0$$

$$\Rightarrow \|\mathbf{y}_t^{pk}\|^2 \mid \mathbf{a} \sim \frac{s_t}{2} \Gamma(N, 2)$$

- $\tilde{\mathbf{y}} = [\|\mathbf{y}_1^{pk}\|^2, \|\mathbf{y}_2^{pk}\|^2, \dots, \|\mathbf{y}_T^{pk}\|^2]^T$
- ML: $\max_{\mathbf{a} \in \{0,1\}^M} p(\tilde{\mathbf{y}}|\mathbf{a}) = \max_{\mathbf{a} \in \{0,1\}^M} \prod_{t=1}^T p(\tilde{y}_t|\mathbf{a})$
$$= \min_{\mathbf{a} \in \{0,1\}^M} \sum_{t=1}^T \left(\frac{\tilde{y}_t}{s_t} + \ln \left(\frac{s_t}{2} \right) \right)$$
s.t. $s_t = P\sigma_h^2 \mathbf{G}_{t:} \text{diag}(\boldsymbol{\beta}) \mathbf{a} + N_0, \forall t$
- Non-convex integer programming problem
- Channel estimation as before

- MAP: Prior $p(\mathbf{a}) = \prod_{m=1}^M p(a_m)$, $p(a_m) = p_a^{a_m} (1 - p_a)^{1 - a_m}$

$$\max_{\mathbf{a} \in \{0,1\}^M} p(\tilde{\mathbf{y}}|\mathbf{a})p(\mathbf{a}) = \max_{\mathbf{a} \in \{0,1\}^M} \left(\prod_{t=1}^T p(\tilde{y}_t|\mathbf{a}) \right) \left(\prod_{m=1}^M p(a_m) \right)$$

$$= \min_{\mathbf{a} \in \{0,1\}^M} \sum_{t=1}^T \left(\frac{\tilde{y}_t}{s_t} + \ln \left(\frac{s_t}{2} \right) \right) + \ln \left(\frac{1 - p_a}{p_a} \right) \sum_{m=1}^M a_m$$

$$\text{s.t. } s_t = P\sigma_h^2 \mathbf{G}_t: \text{diag}(\beta) \mathbf{a} + N_0, \forall t$$

Approach 2:

- τ -length pilots \mathbf{p}_i used with $\|\mathbf{p}_i\|^2 = P$

$$\begin{aligned} \mathbf{Y}_t^{pkH} &= \sum_{i=1}^M g_{ti} a_i \mathbf{p}_i \mathbf{h}_{ti}^H + \mathbf{N}_t^{pH} \\ &= [\mathbf{g}_{t1} \mathbf{p}_1, \dots, \mathbf{g}_{tM} \mathbf{p}_M] \begin{bmatrix} a_1 \mathbf{h}_{t1}^H \\ \vdots \\ a_M \mathbf{h}_{tM}^H \end{bmatrix} + \mathbf{N}_t^{pH} \\ \Rightarrow \underbrace{\mathbf{Y}_t^{pkH}}_{\tau \times N} &= \underbrace{\Phi_t}_{\tau \times M} \underbrace{\mathbf{X}_t}_{M \times N} + \underbrace{\mathbf{N}_t^{pH}}_{\tau \times N} \end{aligned}$$

- Use CS methods to get activity coefficients only
- Discard channel estimates

- Channel estimation:

$$\begin{aligned}
 \mathbf{y}_{tm}^{pk} &= \mathbf{Y}_t^{pk} \mathbf{p}_m = P g_{tm} \hat{a}_m \mathbf{h}_{tm} + \mathbf{N}_t^p \mathbf{p}_m \\
 &\quad + \sum_{i \in \mathcal{S}_k^m} g_{ti} \hat{a}_i \mathbf{h}_{ti} \mathbf{p}_i^H \mathbf{p}_m \\
 \Rightarrow \hat{\mathbf{h}}_{tm}^k &= \mathbb{E} \left[\mathbf{h}_{tm} | \mathbf{y}_{tm}^{pk} \right] \\
 &= \frac{P g_{tm} \hat{a}_m \beta_m \sigma_h^2}{P N_0 + \sigma_h^2 \sum_{i \in \mathcal{S}_k} g_{ti}^2 \hat{a}_i^2 \beta_i |\mathbf{p}_i^H \mathbf{p}_m|^2} \mathbf{Y}_t^{pk} \mathbf{p}_m \\
 \Rightarrow \text{SINR}_{tm}^k &= \frac{P g_{tm}^2 \hat{a}_m^2 \|\hat{\mathbf{h}}_{tm}^k\|^2}{N_0 + P \left(\sum_{i \in \mathcal{S}_k} g_{ti}^2 \hat{a}_i^2 \delta_{ti}^k + \sum_{i \in \mathcal{S}_k^m} g_{ti}^2 \hat{a}_i^2 \frac{|\hat{\mathbf{h}}_{tm}^{kH} \hat{\mathbf{h}}_{ti}^k|^2}{\|\hat{\mathbf{h}}_{tm}^k\|^2} \right)}
 \end{aligned}$$

Approach 3:

- Channel hardening & asymptotic energy based detection

$$\begin{aligned} \frac{\|\mathbf{h}_{tm}\|^2}{N} &\xrightarrow{N \rightarrow \infty} \beta_m \sigma_h^2 \\ \frac{\mathbf{h}_{ti}^H \mathbf{h}_{tj}}{N} &\xrightarrow{N \rightarrow \infty} 0 \\ \implies \frac{\|\mathbf{y}_{tm}^{pk}\|^2}{N} - PN_0 &\xrightarrow{N \rightarrow \infty} \sigma_h^2 \left(\sum_{i \in \mathcal{S}_k} g_{ti}^2 \beta_i a_i^2 |\mathbf{p}_i^H \mathbf{p}_m|^2 \right) =: E_{tm} \end{aligned}$$

- $E_{tm} = \sigma_h^2 \mathbf{G}_t: \text{diag}(\beta) \text{diag}(\mathbf{a}) \mathbf{P}'_{:m}$
- $\mathbf{P}'_{nm} = |\mathbf{p}_n^H \mathbf{p}_m|^2$

- Received energy matrix $\mathbf{E} = \begin{bmatrix} E_{11} & E_{12} & \dots & E_{1M} \\ E_{21} & E_{22} & \dots & E_{2M} \\ \vdots & \vdots & \ddots & \vdots \\ E_{T1} & E_{T2} & \dots & E_{TM} \end{bmatrix}$
- Collect asymptotic energies for all users across all REs

$$E_{tm} = \sigma_h^2 \mathbf{G}_{t:} \text{diag}(\boldsymbol{\beta}) \text{diag}(\mathbf{a}) \mathbf{P}'_{:m}$$

$$\underbrace{\mathbf{E}}_{T \times M} = \underbrace{\sigma_h^2 \mathbf{G}}_{T \times M} \underbrace{\text{diag}(\boldsymbol{\beta})}_{M \times M} \underbrace{\text{diag}(\mathbf{a})}_{M \times M} \underbrace{\mathbf{P}'}_{M \times M}$$

↓ Khatri-Rao vectorization

$$\mathbf{e}_{TM \times 1} = \boldsymbol{\Phi}_{TM \times M} \mathbf{a}'_{M \times 1}$$

References



S. Chen, B. Ren, Q. Gao, S. Kang, S. Sun, and K. Niu, "Pattern division multiple access—a novel nonorthogonal multiple access for fifth-generation radio networks," *IEEE Transactions on Vehicular Technology*, vol. 66, no. 4, pp. 3185–3196, April 2017.



G. Liva, "Graph-based analysis and optimization of contention resolution diversity slotted aloha," *IEEE Transactions on Communications*, vol. 59, no. 2, pp. 477–487, February 2011.



E. Paolini, G. Liva, and M. Chiani, "Coded slotted aloha: A graph-based method for uncoordinated multiple access," *IEEE Transactions on Information Theory*, vol. 61, no. 12, pp. 6815–6832, Dec 2015.



S. Verdu *et al.*, *Multiuser detection*. Cambridge university press, 1998.



K. R. Narayanan and H. D. Pfister, "Iterative collision resolution for slotted aloha: An optimal uncoordinated transmission policy," in *2012 7th International Symposium on Turbo Codes and Iterative Information Processing (ISTC)*, Aug 2012, pp. 136–139.



E. E. Khaleghi, C. Adjih, A. Alloum, and P. Muhlethaler, "Near-far effect on coded slotted aloha," in *2017 IEEE 28th Annual International Symposium on Personal, Indoor, and Mobile Radio Communications (PIMRC)*, Oct 2017, pp. 1–7.



F. Clazzer, E. Paolini, I. Mambelli, and C. Stefanovic, "Irregular repetition slotted aloha over the rayleigh block fading channel with capture," in *2017 IEEE International Conference on Communications (ICC)*, May 2017.



T. L. Marzetta, E. G. Larsson, H. Yang, and H. Q. Ngo, *Fundamentals of massive MIMO*. Cambridge University Press, 2016.