

Coded Power control: Performance Analysis

Benjamin Larrousse and Samson Lasaulce

Laboratoire des Signaux et Systèmes
CNRS - Supélec - Université Paris Sud

July 12th 2013



Outline

- 1 *Introduction: an example*
 - Online Matching Pennies
 - Simple strategies
 - Limiting Performance
 - 2 *General scenario*
 - Problem under study
 - Results
 - 3 *Application*
- References

A Coordination problem: Online Matching Pennies [Gossner et al.]

- Two agents, represented by X_1 and X_2 , with $\mathcal{X}_1 = \mathcal{X}_2 = \{0, 1\}$.
- A random state X_0 i.i.d. $\mathcal{B}(1/2)$, with $\mathcal{X}_0 = \{0, 1\}$.
- Stage utility function:

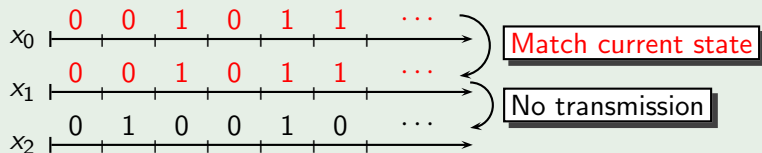
$$w(x_0, x_1, x_2) = \begin{cases} 1 & \text{if } x_0 = x_1 = x_2 \\ 0 & \text{otherwise} \end{cases} .$$

- Particular observation structure: only Agent 1 observes X_0 in advance.
- Scenario repeated T times, $T \in \mathbb{N}$.
- Agents maximize the averaged utility:

$$\underline{w} = \frac{1}{T} \mathbb{E} \left(\sum_{i=1}^T w(x_{0,i}, x_{1,i}, x_{2,i}) \right)$$

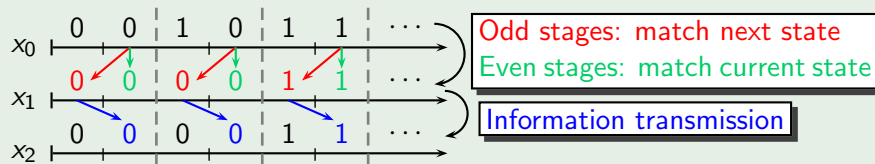
Questions: useful to exchange information ? Maximum utility ?

First Strategy: no communication



→ Expected utility: **0.5**.

Second Strategy: odd/even.



→ Expected utility: **0.625**.

Limiting performance

Result: Online Matching Pennies

- Maximum utility (with global knowledge): 1.
- Maximum with an observation structure (i.e., communication constraint): 0.81.

Questions

- Optimal tradeoff between the costs and the benefits of communication ?
- Characterization of this constraint ?

Problem statement

Definition of the distributed optimization problem

- Two agents X_1, X_2 , a random state X_0 with fixed law $\rho(x_0)$.
- $|\mathcal{X}_i| < \infty$, $i = 0, 1, 2$.
- Utility over T stages:

$$\underline{w} = \frac{1}{T} \mathbb{E} \left(\sum_{i=1}^T w(x_{0,i}, x_{1,i}, x_{2,i}) \right)$$

- Y signal on X_1 with fixed law $\Gamma(y|x_1)$.
- Strategy of Agent 1: $(\sigma_{1,i})_{1 \leq i \leq T}$ with:

$$\sigma_{1,i} : \mathcal{X}_0^T \times \mathcal{X}_1^{i-1} \times \emptyset \rightarrow \mathcal{X}_1$$

- Strategy of Agent 2: $(\sigma_{2,i})_{1 \leq i \leq T}$ with:

$$\sigma_{2,i} : \mathcal{X}_0^{i-1} \times \mathcal{Y}^{i-1} \times \mathcal{X}_2^{i-1} \rightarrow \mathcal{X}_2$$

Auxiliary notion

Definition (Implementability)

$P_{X_{0,i}, X_{1,i}, X_{2,i}, Y_i}$: joint distribution induced by $(\sigma_{1,i}, \sigma_{2,i})_{i \geq 1}$ at stage i .
 The distribution $\bar{Q}(x_0, x_1, x_2)$ is implementable if there exists a pair of strategies $(\sigma_{1,i}, \sigma_{2,i})_{i \geq 1}$ such that for all (x_0, x_1, x_2) ,

$$\frac{1}{T} \sum_{i=1}^T \sum_y P_{X_{0,i}, X_{1,i}, X_{2,i}, Y_i}(x_0, x_1, x_2, y) \rightarrow \bar{Q}(x_0, x_1, x_2)$$

as $T \rightarrow +\infty$.

Feasible utilities

A certain utility value \underline{w} is reachable if and only if there exists an implementable distribution Q such that $\underline{w} = \mathbf{E}_Q[w]$.

Theorem (Characterization of implementable distribution)

Let $\bar{Q} \in \Delta(\mathcal{X}_0 \times \mathcal{X}_1 \times \mathcal{X}_2)$ with $\sum_{(x_1, x_2)} \bar{Q}(x_0, x_1, x_2) = \rho(x_0)$. The distribution \bar{Q} is implementable if and only if there exists $Q \in \Delta(\mathcal{X}_0 \times \mathcal{X}_1 \times \mathcal{X}_2 \times \mathcal{Y})$ which verifies:

$$I_Q(X_0; X_2) \leq I_Q(X_1; Y | X_0, X_2)$$

where the arguments of the mutual information $I_Q(\cdot)$ are defined from Q and $Q(x_0, x_1, x_2, y) = \bar{Q}(x_0, x_1, x_2)\Gamma(y|x_1)$.

Remark: This theorem also characterizes expected payoff.

Optimization problem

Our problem is an optimization problem with *convex* constraint:

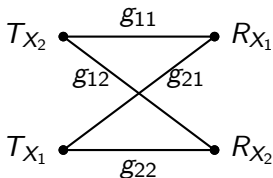
$$\begin{aligned}
 &\text{minimize} && -\mathbb{E}_{\underline{q}}[w] = -\sum_{\ell=1}^L q_{\ell} w_{\ell} \\
 &\text{subject to} && I_{\underline{q}}(X_0; X_2) - I_{\underline{q}}(X_1; Y|X_0, X_2) \leq 0 \\
 & && -q_{\ell} \leq 0 \\
 & && -1 + \sum_{\ell=1}^L q_{\ell} = 0 \\
 & && \forall x_0, \sum_{\ell \in \mathcal{L}_{X_0}(x_0)} q_{\ell} - \rho(x_0) = 0 \\
 & && \forall (x_1, y), \frac{\sum_{\ell \in \mathcal{L}_{X_1, Y}(x_1, y)} q_{\ell}}{\sum_{\ell \in \mathcal{L}_{X_1}(x_1)} q_{\ell}} - \Gamma(y|x_1) = 0
 \end{aligned}$$

It can be solved numerically.

Application: flat-fading interference channel (IC)

Application: power control

- Agents: transmitters.
- Actions: power levels.
- Random state: channel gains.



$$\forall i, k \in \{1, 2\} : g_{ik} \in \{0.1, 1.9\}$$

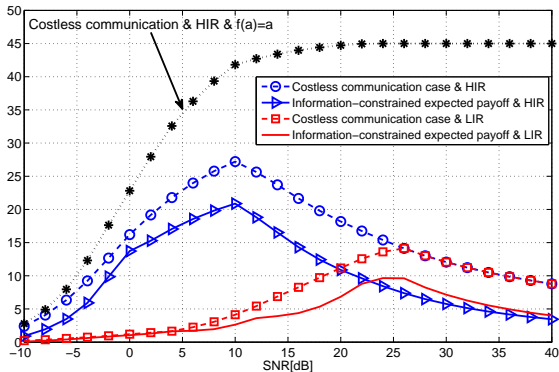
$$g_{ik} \sim \mathcal{B}(\pi_{ik})$$

$$x_i \in \mathcal{X}_i = \{0, P_{\max}\}$$

- Utility: $w^{\text{IC}} = \sum_{i=1}^2 \log_2(1 + \text{SINR}_i)$ with $\text{SINR}_i = \frac{g_{ii}x_i}{\sigma^2 + g_{ki}x_k}$.
- Observation: $Y \equiv X_1$.

Coded Power Control policies :

- ▶ Semi-coordinated PC (SPC) policy: $x_2 = P_{\max}$, $x_1^\dagger \in \arg \max_{x_1} w^{\text{IC}}(x_0, x_1, P_{\max})$.
- ▶ Optimal CPC policy: maximum of the convex optimization problem.
- ▶ Costless communication case: maximum of w^{IC} at any stage.






Conclusion

- Shannon theory helps us to find the right information constraint.
- Explicit characterization of feasible utility.
- Methodology used to derive optimal performance is general.

Thank you for your attention.
Do not hesitate to ask questions.
For more: <http://benjamin.larrousse.fr/>

References

-  Cover, T. M., and Thomas, J. A. (2006). *Elements of information theory, 2nd Edition*. Wiley Series in Telecommunications. Wiley.
-  Gossner O., Hernandez P., Neyman A. (2006) *Optimal use of communication resources*, *Econometrica*, Vol.74, No.6, 1603-1636.
-  Lasaulce S. and Tembine H. (2011) *Game Theory and Learning for Wireless Networks : Fundamentals and Applications*. Academic Press, 2011.