CS 6840 – Algorithmic Game Theory

Spring 2012

Lecture 14 – Wednesday 22 February 2012 – Scribe Notes

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## Coarse correlated equilibria as a convex set

Last time - Looked at algorithm that guarantees no regret Last last time - Defined coarse correlated equilibrium as a probability distribution on strategy vectors

**Definition.** p(s) s.t.  $E(u_i(s)) \ge E(u_i(x, s_{-i})) \ \forall i, \forall x.$ 

which lead to the corollary:

Corollary 1. All players using small regret strategies gives an outcome that is close to a coarse correlated equilibrium

The next natural question to ask is: Does there exist a coarse correlated equilibrium? We consider finite player and strategy sets.

**Theorem 2.** With finite player and strategy sets, a coarse correlated equilibrium exists.

*Proof 1.* We know that a Nash equilibrium exists. Then let  $p_1, \ldots p_n$  be probability distributions that form a Nash equilibrium. Observe that  $p(s) = \prod_i p_i(s_i)$  is a coarse correlated equilibrium.

Proof 2. (doesn't depend on Nash's theorem). Idea: Algorithm from last lecture finds it with small error. Consider

$$\min_{p}[\max_{i}[\max_{x}[E_{p}(u_{i}(x_{i},s_{-i}))-E_{p}(u_{i}(s))]]]$$

The quantity inside the innermost max is the regret of players i about strategy x. If this minimum is  $\leq 0$ , then p is a coarse correlated equilibrium. The minimum cannot equal  $\epsilon > 0$  as we know by the algorithm that we can find a p with arbitrarily small regret. In this instance,  $\frac{\epsilon}{2}$  would be sufficient to reach a contradiction. Hence, we know that the infimum must be less than or equal to 0 but does the minimum exist? Since we have a continuous function over p, the compact space of probability distributions, we must attain the infimum, so the minimum is in fact  $\leq 0$ , so a coarse correlated equilibrium exists.

**Remark.** This minimum can be calculated as the solution of a linear program satisfying  $\sum p(s) = 1, p(s) \ge 0$  and the no regret inequality for each (i, x) pair.

## 2-person 0-sum games

The game is defined by a matrix a with the first players strategies labelling the rows and the second players strategies labelling the columns.  $a_{ij}$  is the amount Player 1 pays to Player 2 if strategy vector (i, j) plays.

**Theorem 3.** Coarse correlated equilibrium in these games is (essentially) the same as the Nash equilibrium.

To be a bit more precise, let p(i,j) be at coarse correlated equilibrium. When considering Player 1, we care about q Player 2's marginal distribution.  $q(j) = \sum_i p(i,j)$ . Since Player 1 has no regret, we have that

$$\sum_{ij} a_{ij} p(i,j) \le \min_{i} \sum_{j} a_{ij} q_{j}$$

Likewise, let  $r(i) = \sum_j p(i,j)$  be Player 1's marginal distribution, so Player 2's lack of regret tells that:

$$\sum_{ij} a_{ij} p(i,j) \ge \max_{j} \sum_{i} a_{ij} r_{i}$$

**Theorem 4.** q, r from above are Nash equilibria.

*Proof.* The best response to q is

$$\min_{i} \sum_{j} a_{ij} q_{j} \le \sum_{ij} r(i) q(j) \le \max_{j} \sum_{i} a_{ij} r_{i}$$

the last of which is the best response to r. Thus, we also have

$$\sum_{ij} a_{ij} p(i,j) \le \min_{i} \sum_{j} a_{ij} q_j \le \sum_{ij} r(i) q(j) \le \max_{j} \sum_{i} a_{ij} r_i \le \sum_{ij} a_{ij} p(i,j)$$

Which implies the result, since they must all be equal.