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## Mean-Payoff games

### Lecture Notes - II

#### Organization:

1. Recall previous results
2. Understanding finite tree games; ~~per~~ effects of perturbing the values
3. Intermediate game; proof of equivalence of values with std. mean-payoff
4. Proof of positional determinacy
5. Computing values of all vertices uniformly
6. Uniform positional strategies.

#### Part 1: Recall previous results

Mean-Payoff game (infinite)  $\rightarrow G$

First-cycle mean-payoff  $\rightarrow F$

Both start from specified vertex  $a_0$ .

(Infinite) Result: There exist numbers  $\tilde{v}(a_0), \tilde{\sigma}, \tilde{\tau}$  s.t. when ~~Max~~

in  $G$ :  $\forall \tau \text{ Payoff}(\tilde{\sigma}, \tau) \geq \tilde{v}(a_0)$   
 $\forall \sigma \text{ Payoff}(\sigma, \tilde{\tau}) \leq \tilde{v}(a_0)$

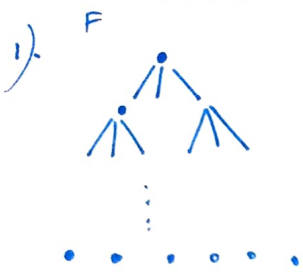
$\tilde{v}(a_0)$  is minimax equilibrium, called value of the game.

(Finite) Result: Similar  $v(a_0), \sigma, \tau$  for first-cycle version.

Transfer result:  $v(a_0) = \tilde{v}(a_0)$ .  $\tilde{\sigma}$  can be obtained from  $\sigma$  by the stack-based method.  
 $\tilde{\tau}$  can be obtained from  $\tau$

To show:  ~~$\exists$~~   $\exists$  there exist positional strategies giving the optimal value.

Part 2: Some properties of the finite tree games



Each leaf has a payoff.  
Min-Max computation gives  $v, \sigma, \tau$

- 2) Restrict the tree to  $\sigma$
- Root has value  $v$ .  
if it is min vertex, every child has value  $\geq v$ .
  - if max vertex, then the single edge out of it is given by  $\sigma$ , and value equals max

3) Inductively we can show that all labels in  $F_\sigma$  have  $\geq v$ .

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Similarly for  $F_\tau$  we can show all values  $\leq v$ .

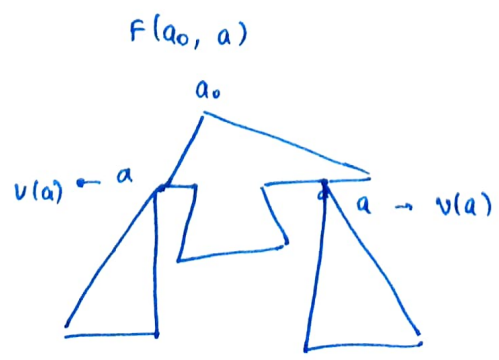
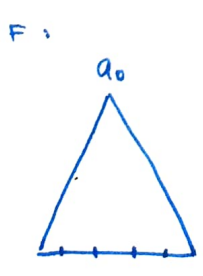
Exercise: Suppose some leaves of  $F$  have ~~values~~ <sup>payoffs</sup> changed to  $v$ . Will the value of the game change?

Proof: No. Consider same strategies  $\sigma, \tau$ . In  $F_\sigma$ , all leaves have  $\geq v$  so playing  $\sigma$  ensures  $\geq v$ . Similarly playing  $\tau \leq v$ . Value of the game is unique. No other no. can have this property. So, modifying with  $v$  does not affect the value.

Part 3: An intermediate game.

$(G, a_0)$   $(F, a_0)$ .  $(F, a_0, a) \rightarrow$  new game.

$(F, a_0, a)$ : ~~pro~~ starts at  $a_0$  and proceeds like  $F$ . If  $a$  does not come in the play, then it stops when there is a repetition (identical to  $F$ ). When there is an 'a', the game  $F$  starts from 'a'. Essentially the prefix till now is forgotten and a new first cycle game started from  $a$ .



What is value of  $(F, a_0, a)$ ?

③



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Claim 1: Let  $\sigma$  be optimal strategy for Max in  $F$ .  
If 'a' occurs in  $F_\sigma$ , then  $v(a) \geq v(a_0)$ .

Proof: Pitfall: ~~Just~~ the min-max label on 'a' in  $F_\sigma$  need not be  $v(a)$ . So this is not immediate.

We will use the infinite version for this and show that  $\tilde{v}(a) \geq \tilde{v}(a_0)$ .

Consider the stack strategy  $\tilde{\sigma}(a_0)$ , and consider the infinite tree  $G_{\tilde{\sigma}}$



We have seen that every play following  $\tilde{\sigma}$  has value  $\geq \tilde{v}(a_0)$ .

- a is a node in this tree.
- a occurs at multiple positions.
- Pick one occurrence of 'a' and look at the infinite tree below it.

- The tree below 'a' gives a strategy for ~~Max~~ for the game starting at 'a'. ~~By playing~~ call this  $\tilde{\sigma}_a$ . ~~By playing~~ since payoffs are not dependent on the prefix, each play starting from 'a' has the same payoff  $\geq \tilde{v}(a_0)$ . So

~~Similarly, we can look at  $G_{\tilde{\sigma}}$~~

Therefore, by playing this strategy, Max can ensure at least  $\tilde{v}(a_0)$ .

$\Rightarrow$  optimal strategy for Max for game starting at 'a' is at least

$\tilde{v}(a_0)$ .

$$\Rightarrow \tilde{v}(a) \geq \tilde{v}(a_0)$$

$$\Rightarrow v(a) \geq v(a_0)$$

(4)

Claim 2: Let  $\tau$  be optimal strategy for Min in  $F$ . If 'a' occurs in  $F_\tau$ , then  $v(a) \leq \cancel{v(a_0)}$ .  $v(a_0)$

Similar proof as before.

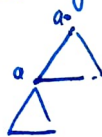
Claim 3: Consider game  $(F, a_0, a)$ . Value of this game equals  $v(a_0)$ .

Let  $\mu(a_0)$  be the value.

Consider optimal strategy  $\sigma_1$  in  $F$ . Suppose 'a' never occurs in the tree  $F_\sigma$ , then play  $\sigma$  will result in a game similar to  $F$ . So  $\mu(a_0) = v(a_0)$

However, suppose 'a' occurs in  $F_\sigma$ . Then from Claim 1,  $v(a) \geq v(a_0)$

~~Hence (from discussion)~~ The tree starting at 'a' will have value  $v(a)$ .



- This from the discussion in Part 2, we get that  $\mu(a_0) \geq v(a_0)$ .

- Similarly, when Min plays  $\tau$ , by Claim 2, we can show that  $\mu(a_0) \leq v(a_0)$ .

Result: Value of the intermediate game  $\mu(a_0) = v(a_0)$ .



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Part 4: Proof of positional determinacy

Game starts at  $a_0$ . We will show positional strategies

$\sigma, \tau$  for  $G$  and  $F$ .

~~Proof~~ First we show positional strategy  $\sigma$  for Max giving optimal value.

Proof is by induction on  $|E| - |V_{\max}|$ . We assume  $|E| - |V_{\max}| \geq 0$ .  
From every vertex, at least one outgoing edge.

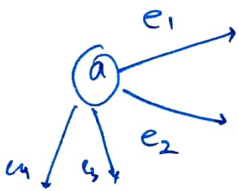
Base case:  $|E| - |V_{\max}| = 0$

Every Max strategy is positional. So done.

Induction case: Assume  $|E| - |V_{\max}| = k+1$ , and the thm. true for

$|E| - |V_{\max}| = k$ .

Pick a vertex  $a'$  <sup>of Max</sup> s.t. there are at least 2 outgoing edges.

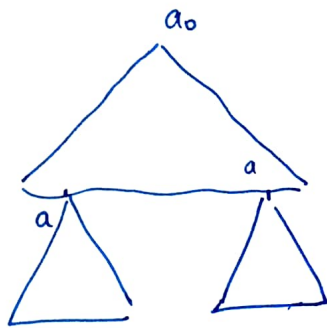


$G_1$ : ~~restrict G to~~  $e_1$

Consider game ~~to~~  $(F, a_0, a)$  starting from  $a_0$ .

consider optimal strategy  $\sigma_a$  for Max

We can assume that Max plays the same way in all trees starting with 'a'.



Moreover 'a' occurs only once in a non-leaf node. Let 'e1' be the edge played ~~to~~ from 'a'. Restrict the graph to

'e1' alone out of 'a'. Value  $v(a_0)$  will not change in this graph, because  $\mu(a_0)$  does not change!

By induction <sup>hyp.</sup> there is a positional strategy <sup>for Max</sup> for this game. The same strategy works in the bigger game ( $\geq v(a_0)$ )

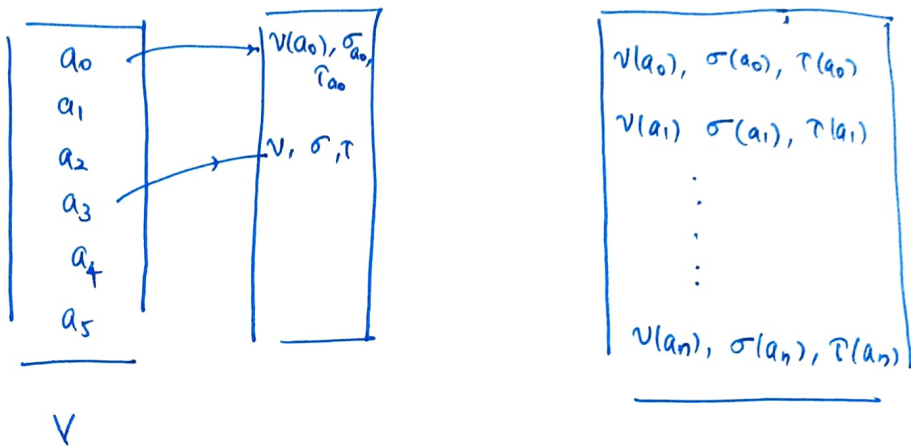
Similar proof shows positional strategy for Min.

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Part 5: Computing values of vertices uniformly for all vertices.

In the previous discussion, we have considered the game starting at  $a_0$ . We have shown that there exist positional strategies  $\tilde{\sigma}, \tilde{\tau}$  that achieve  $\hat{v}(a_0)$ .

This gives a map:



Questions: 1. Can we compute  $\langle v(a_0), \dots, v(a_n) \rangle$  all together?

2. Do there exist uniform positional strategies  $\sigma, \tau$  that work for every vertex, and give the required value?

Uri-Zwick's k-step approximation procedure:

So far we have looked at values for games starting from a specific vertex. To get a global picture, we need to be able to relate value of a vertex to values of its successors.

→ Consider a game starting from 'a' and which goes for  $k$ -rounds.

The payoff after  $k$ -rounds is the sum of weights on the  $k$ -edges seen. Min wants to minimize this and Max wants to maximize this.

→ Since this is a finite <sup>duration</sup> game, values exist by the min-max

computation. Let  $v_k(a)$  denote value of  $k$ -step game starting from 'a'.