# Computing Equilibria of Repeated And Dynamic Games

Şevin Yeltekin

Carnegie Mellon University

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# **DYNAMIC GAMES**

## A specific example: Dynamic Oligopoly

Oligopoly game with endogenous productive capacity.

- Study the nature of dynamic competition and its evolution.
- Study the nature of cooperation and competition.
- Specifically:
  - Is ability to collude affected by state variables?
  - Do investment decisions increase gains from cooperation?
  - Does investment present opportunities to deviate from collusive agreements?

## Existing Literature in Dynamic Oligopoly

#### Existing literature in IO

- Two stage games
  - Firms choose capacities in stage one, prices in stage two
  - Kreps-Scheinkman (1983), Davidson-Deneckere (1986)
- Dynamic games
  - Firms choose capacities and prices
  - Benoit-Krishna (1987), Davidson-Deneckere (1990)

#### Goals revisited

- Limiting assumptions in previous work
  - Capacity chosen at t=0 , OR
  - No disinvestment, OR
  - Examine only equilibria supported by Nash reversion, OR
  - Restrictive functional forms for demand and cost functions
- Our goal: Examine full set of pure strategy Nash equilibria for dynamic games with arbitrary cost and demand functions.

## Stage Game: Environment

- N infinitely lived agents.
- Individual state:  $x_i \in X_i$
- Aggregate state:  $x \in X = \times_{i=1}^{N} X_i$
- Finite action space for player i:  $A_i$ , i = 1, ..., N
- Action profiles:  $A = \times_{i=1}^{N} A_i$
- Aggregate state evolution:  $g: A \times X \to X$

## Stage Game: Payoffs

- Per period payoff function  $\Pi_i:A\to\Re$
- Minimal payoffs

$$\underline{\Pi}_{i,x} \equiv \min_{a \in A} \Pi_i(a,x)$$

Maximal payoffs

$$\overline{\Pi}_{i,x} \equiv \max_{a \in A} \Pi_i(a,x)$$

ullet Equilibrium payoffs in state x contained in

$$W_x = \times_{i=1}^N [\underline{\Pi}_{i,x}, \overline{\Pi}_{i,x}].$$

• Payoff correspondence:

$$W:X \rightrightarrows \Re^N$$

### Dynamic Game

- Action space:  $A^{\infty}$
- $h_t$ : t-period history:

$$\{\{a_s, x_s\}_{s=0}^{t-1}, x_t\}$$
 with  $x_s = g(x_{s-1}, a_{s-1}), a_s \in A$ 

- Set of t-period histories:  $H_t$
- Preferences:

$$w_i(a^{\infty}, x^{\infty}) = \frac{1 - \delta}{\delta} E_0 \sum_{t=1}^{\infty} \delta^t \Pi_i(a_t, x_t).$$

• Strategies:  $\{\sigma_{i,t}\}_{t=0}^{\infty}$  with  $\sigma_{i,t}: H_t \to A_i$ .

### Equilibrium Payoff Correspondence

- SPE payoff correspondence:  $V^* \equiv \{V_x^* | x \in X\}$
- $\mathcal{P}$ : set of all correspondences  $\mathcal{W}: X \rightrightarrows \Re^N$  s.t.
  - ullet Graph of  ${\mathcal W}$  is compact
  - Graph of  ${\mathcal W}$  contained within Graph of  ${\mathcal P}.$
  - $V^*$  may be shown to be an element of  $\mathcal{P}$ .

### Steps: Computing the Equilibrium Value Correspondence

- 1 Define an operator that maps today's equilibrium values to tomorrow's at each state.
- Show that this operator is monotone and the equilibrium correspondence is its largest fixed point.
- 3 Define approximation for operator and correspondences that
  - Represents correspondence parsimoniously on computer
  - Preserves monotonicity of operator
- 4 Define an appropriately chosen initial correspondence, apply the monotone operator until convergence.

# Step 1: Set Valued Dynamic Programming

- Recursive formulation
- Each SPE payoff vector is supported by
  - profile of actions consistent with Nash today
  - continuation payoffs that are SPE payoffs
- ullet Construct self-generating correspondences to find  $V^*$

### Step 1: Operator

$$B^*: \mathcal{P} \to \mathcal{P}$$
.

• Let  $\mathcal{W} \in \mathcal{P}$ .

$$B^*(\mathcal{W})_x = \bigcup_{(a,w)} \{ (1-\delta)\Pi(a,x) + \delta w \}$$

subject to:

$$w \in \mathcal{W}_{q(a,x)}$$

and for each  $\forall i \in N, \, \forall \tilde{a} \in A_i$ 

$$(1 - \delta)\Pi_i(a, x) + \delta w_i \ge \Pi_i(\tilde{a}, a_{-i}, x) + \delta \mu_{i, g(\tilde{a}, a_{-i}, x)}\}$$

where  $\mu_{i,x} = \min\{w_i | w \in \mathcal{W}_x\}.$ 

## Step 2: Self-generation

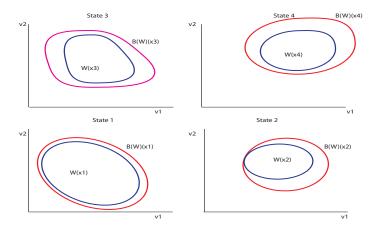
A correspondence W is self-generating if :

$$\mathcal{W} \subseteq B^*(\mathcal{W}).$$

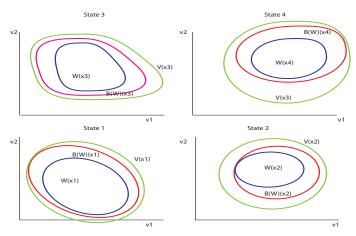
An extension of the arguments in APS establishes the following:

- Graph of any self-generating correspondence is contained within  $Graph(V^*)$ ,
- $V^*$  itself is self-generating.

# Self-generation visually



# Self-generation visually

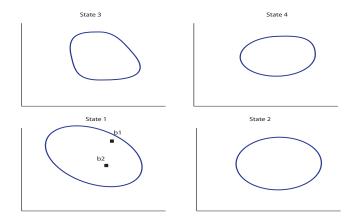


#### Step 2: Factorization

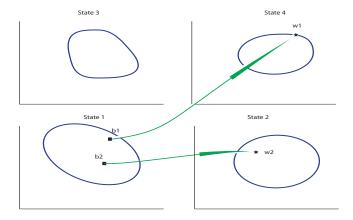
 $b \in B^*(\mathcal{W})_x$  if there is an action profile a and continuation payoff  $w \in \mathcal{W}_{g(a,x)}$ , s.t

- b is value of playing a today in state x and receiving continuation value w ,
- for each i, player i will choose to play  $a_i$
- $x\prime = g(a,x)$  if no defection
- $\tilde{x} = g(\tilde{a}_i, a_{-i}, x)$  if defection.
- punishment value drawn from set  $\mathcal{W}_{\widetilde{x}}$ .

#### Factorization I



#### Factorization II



## Step 2: Eqm Value Correspondence as Fixed Point

• Monotonicity:  $B^*$  is monotone in the set inclusion ordering:

If 
$$W_1 \subseteq W_2$$
, then  $B^*(W_1) \subseteq B^*(W_2)$ 

- Compactness:  $B^*$  preserves compactness.
- Implications:
  - 1)  $V^*$  is the maximal fixed point of the mapping  $B^*$ ;
  - 2)  $V^*$  can be obtained by repeatedly applying  $B^*$  to any set that contains graph of  $V^*$ .

## Step 3: Approximating Value Correspondences

- Represent candidate value correspondences on computer
- Preserve monotonicity of operator
- Proceed in 2 steps
  - 1 Convexify underlying game.
  - 2 Develop method for approximating convex-valued correspondences.

### Step A: Public randomization

- Public lottery with support contained in  $\mathcal{W}_{g(a,x)}$ .
- Public lottery specifies continuation values for the next period
  - Lottery dependent on current actions determines Nash equilibrium for next period.
  - Strategies now condition on histories of actions and lottery outcomes.
- Modified operator:

$$B(W) = co(B^*(co(W))), \qquad W \in \mathcal{P}.$$

- V equilibrium value correspondence of supergame with public randomization.
- B is monotone and V is the largest fixed point of B.

## **Environment: Dynamic Cournot with Capacity**

- Firm i has sales of  $q_i \in Q_i(k_i)$ , and unit cost  $c_i$ .
- MC= maintenance cost of machine
- SP= resale/scrap value of machine
- FC =cost of a new machine
- Cost of capital maintenance and investment:

$$C(k_i, k_i') = \begin{cases} MC * (k_i - 1) + FC * (k_i' - k_i) & \text{if } k_i' \ge k_i \\ MC * (k_i - 1) - SP * (k_i - k_i') & \text{if } k_i' \le k_i \end{cases}$$

## Profit: Dynamic Cournot with Capacity

• Firm *i*'s current profits:

$$\Pi_i(q_1, q_2, k_i, k_i') = q_i(p(q_1, q_2) - c_i) - C(k_i, k_i')$$

Linear demand curve:

$$p(q_1, q_2) = \max \{a - b(q_1 + q_2), 0\}.$$

## Stage Game: Dynamic Cournot with Capacity

- Action Space:
  - sets of outputs
  - sets of capital stocks
- State Space:
  - set of feasible capital stocks
- $A_i = Q_i \times K_i$
- $X = K_1 \times K_2$

### Dynamic Strategies and Payoffs

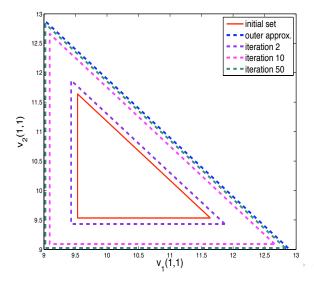
- Strategies: collection of functions that map from histories of outputs and capital stocks into current output and capital choices.
- Maximize average discounted profits.

$$\frac{(1-\delta)}{\delta} \sum_{t=0}^{t=\infty} \delta^t \Pi_{i,t}(q_1, q_2, k_i, k_i')$$

## Dynamic Duopoly: Example 1

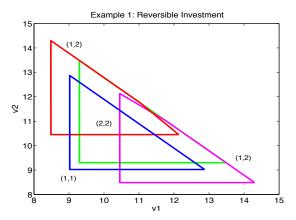
- Finite action version of the dynamic duopoly game.
- Discretize action space over  $q_i$  and  $k_i$
- ullet Full capacity: Actions from interval  $[0,ar{Q}]$
- Partial capacity: Actions from interval  $[0, \bar{Q}/2]$
- Firms endowed with 1 machine each.
- 4 states:  $(k_1, k_2) \in \{(1, 1), (1, 2), (2, 1), (2, 2)\}$
- 48 hyperplanes for the approximation.

## Monotone Operator and Convergence

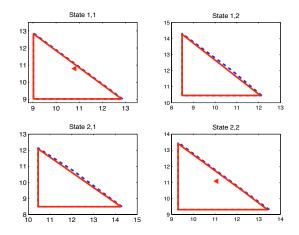


#### Fluctuation Market Power

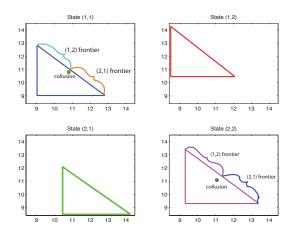
Parameters: MC =SP=1.5, FC =2.5,  $\delta=0.8$ ,  $\bar{Q}=6.0$  c=0.6, b=0.3, a=6.0  $p(q_1,q_2) = \max{\{a-b(q_1+q_2),0\}}.$ 



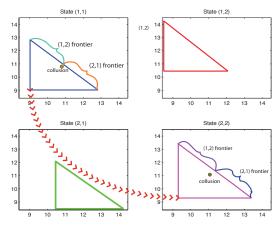
#### **Error Bounds**



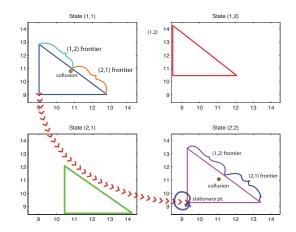
## Striving for Market Power I



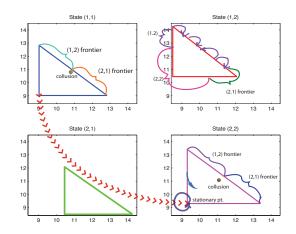
## Striving for Market Power II



## Striving for Market Power III



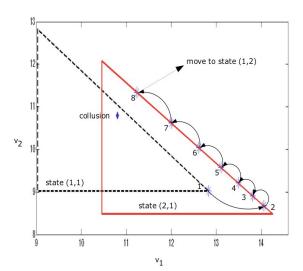
## Striving for Market Power : Strategies



### Strategies: Fluctuating Market Power

- Firms can do better than symmetric Nash collusion.
- Frontier of equilibrium value sets supported by
  - continuation play where firms alternate having market power.
- Worst equilibrium payoffs
  - · firms produce at full capacity in current period
  - over-investment and over-production thereafter (symmetric cases).

## Striving for Market Power: Strategies

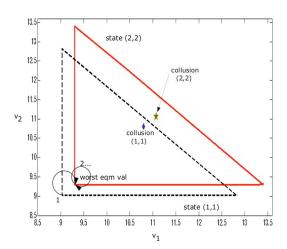


## Striving for Market Power : Strategies

Table: Equilibrium Path

Node	$v_1$	$v_2$	$k_1$	$k_2$	$q_1$	$q_2$
1	12.8289	9.0232	1	1	3.0	3.0
2	14.0571	8.6750	2	1	6.0	3.0
3	13.8064	8.9118	2	1	6.0	3.0
4	13.4930	9.2078	2	1	6.0	3.0
5	13.1012	9.5777	2	1	6.0	3.0
6	12.6115	10.0401	2	1	6.0	3.0
7	11.9994	10.6181	2	1	6.0	3.0
8	11.2342	11.3407	2	1	6.0	3.0

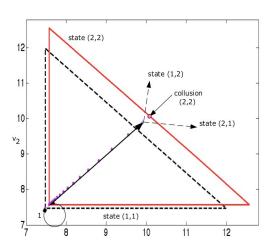
### Worst Equilibrium



#### Worst Equilibrium c=0.9

- With higher per unit cost (c=0.9), playing uncooperatively too costly.
- Following one period of over investment and over production
  - Firms move towards Pareto frontier.
  - Continuation values increasing over time
  - Followed by alternating market power and high profits
- Nature of cooperation depends on state and on history.
- Markov perfect eqm. cannot capture this.

## Striving for cooperation



# Striving for cooperation

Node	$v_1$	$v_2$	$k_1$	$k_2$	$q_1$	$q_2$
1	7.47280025107915	7.47280025107915	1	1	3.0	3.0
2	7.57200031384894	7.57200031384894	2	2	6.0	6.0
3	7.59000039231118	7.59000039231118	2	2	6.0	6.0
4	7.61250049038897	7.61250049038897	2	2	6.0	6.0
5	7.64062561298621	7.64062561298621	2	2	6.0	6.0
6	7.67578201623276	7.67578201623276	2	2	6.0	6.0
7	7.71972752029095	7.71972752029095	2	2	6.0	6.0
8	7.77465940036369	7.77465940036369	2	2	6.0	6.0
9	7.84332425045461	7.84332425045461	2	2	6.0	6.0
10	7.92915531306827	7.92915531306827	2	2	6.0	6.0
11	8.03644414133533	8.03644414133533	2	2	6.0	6.0
12	8.17055517666916	8.17055517666916	2	2	6.0	6.0
13	8.33819397083645	8.33819397083645	2	2	6.0	6.0
14	8.54774246354557	8.54774246354557	2	2	6.0	6.0
15	8.80967807943196	8.80967807943196	2	2	6.0	6.0
16	9.13709759928994	9.13709759928994	2	2	6.0	6.0
17	9.54754361279848	9.54754361279842	2	2	6.0	6.0
18	10.0594295159981	10.0594295159980	2	1		
			1	2		

#### Summary

- Computation of equilibrium value correspondence reveals
  - dynamic interaction and competition missed by simplifying assumptions
  - rich set of equilibrium outcomes that involve
    - fluctuating market power
    - over-investment and over-production when cooperation breaks down
    - · worst equilibrium resembles prisoner's dilemma
    - best equilibria resemble battle of the sexes.
    - equilibria with current profit of leading firm less than smaller firm

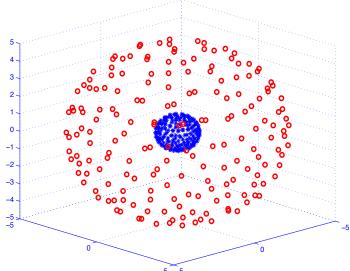
#### Supergames with Continuous States

- Approximation substantially more complicated than discrete states.
- Goal: Find an approximation scheme with right properties that preserves outer/inner bounds.
- Use set-valued step functions.
- See unpublished mimeo: Sleet and Yeltekin (1999); "On the approximation of value correspondences".

## Number of players

- So far examples have N=2.
- Algorithm applicable to  ${\cal N}>2$
- Some computational issues.
  - Computational power. No of optimizations rise exponentially.
  - Choice of hyperplanes non-trivial. [Sampling on a sphere.]
  - Harder to define/calculate error bounds.

# Sampling surface of sphere



#### Continuous Actions

- Optimizations are LP problems.
- LP has nearly negligible approximation error.
- Using LP ensures outer and inner approx. do not have optimization error.
- NLP methods can introduce optimization errors that distort the inner/outer structure.
- My advice: Stick to discrete actions.

## Example: Behavioral Economics Applied to Poverty

- Bernheim, Ray, Yeltekin (2013), "Poverty and Self Control"
- intertemporal allocation problem with credit constraints faced by an individual with quasi-hyperbolic preferences
- use method to study all SPE
- show that there is a poverty trap: no personal rule permits the individual to avoid depleting all liquid wealth. Poverty perpetuates itself by undermining the ability to exercise self-control.

# Example: Dynamic Games in Macro Policy Making

- Credible policy designed as dynamic game between planner +continuum of agents with capital.
- One large strategic player + continuum of non-strategic players.
- How does one apply a variant of APS?
- Use planner's value and tomorrow's marginal utility of capital.

## Examples: Dynamic Games in Macro Policy Making

- Optimal Fiscal Policy in a Business Cycle Model without Commitment (Fernandez-Villaverde, Tsyvinski, 2002)
  - Use method to characterize Sustainable/Credible Equilibria.
    Compute eqm strategies and calibrate data to the US.
- On Credible Monetary Policy and Private Government Information (Chris Sleet, JET, 2001).
- Phelan and Stacchetti (Econometrica, 2001): Ramsey tax model w/ capital and no govt commitment.
  - Use planner's value and tomorrow's marginal utility of capital.