# Solving Dynamic Games with Newton's Method 

Michael C. Ferris<br>University of Wisconsin

Kenneth L. Judd

Hoover Institution

Karl Schmedders<br>Kellogg School of Management

2006 Institute on Computational Economics
Argonne National Laboratory
July 21, 2006

## Motivation

Stochastic, finite-state dynamic games have many applications in economics

Arise frequently in imperfect competition models

- Merger analysis (Gowrisankaran, 1999)
- Learning by doing (Benkard, 2000)
- Collusion (Fershtman and Pakes, 2000, de Roos, 2004)
- Capacity games (Besanko and Doraszelski, 2004)
- Advertising (Doraszelski and Markovich, 2005)


## Solving Interesting Models

Numerical methods needed for solving non-trivial models

- Pakes and McGuire (1994, 2001)
- Doraszelski and Judd (2005)

Problem: Computational costs restrict applications

Our paper: We propose a simple method for solving large models

## Overview of this Talk

- Description of general discrete-time stochastic games
- Basic idea of existing methods
- Newton method
- Application: Two-firm example with investment and production
- Conclusion: It is feasible to solve large games


## Discrete-Time Dynamic Game

Stochastic discrete-time dynamic game (for two players)

State variables

- Represent production capacity, efficiency, experience, etc.
- State of firm $i$ at time $t$ is $\omega_{t}^{i}$
- State of game is $\omega_{t}=\left(\omega_{t}^{1}, \omega_{t}^{2}\right) \in \Omega$


## Actions

- Represent output, price decision, investments, etc.
- Firm $i$ 's action at time $t$ is $x_{t}^{i} \geq 0$
- Collection of actions at $t$ is $x_{t}=\left(x_{t}^{1}, x_{t}^{2}\right)$


## Discrete-Time Dynamic Game II

## Stochastic process of state-to-state transitions

- Represents uncertainty about investment success, depreciation, etc.
- Transition probabilities

$$
\operatorname{Pr}\left(\omega_{t+1}=\xi \mid \omega_{t}, x\right)=\operatorname{Pr}^{1}\left(\omega_{t+1}^{1}=\xi^{1} \mid \omega_{t}^{1}, x_{t}^{1}\right) \cdot \operatorname{Pr}^{2}\left(\omega_{t+1}^{2}=\xi^{2} \mid \omega_{t}^{2}, x_{t}^{2}\right) .
$$

- Independent transitions, each firm controls its state


## Payoffs

- Represent net profits from current sales, investment expenditures, etc.
- Firm $i$ receives $\pi^{i}\left(x_{t}, \omega_{t}\right)$ at time $t$


## Discrete-Time Dynamic Game III

Objective functions

- Represent total profit over an infinite horizon

$$
E\left\{\sum_{t=0}^{\infty} \beta^{t} \pi^{i}\left(x_{t}, \omega_{t}\right)\right\}
$$

- Both firms simultaneously maximize respective total profits


## Pure Markov Strategies

Firm $i$ uses a strategy of feedback form, $X^{i}(\omega)$
Firm $i$ 's expected net present value $V^{i}(\omega)$
Bellman equations for the two firms

$$
\begin{aligned}
& V^{1}(\omega)=\max _{x^{1}} \pi^{1}\left(x^{1}, X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, x^{1}, X^{2}(\omega)\right\} \\
& V^{2}(\omega)=\max _{x^{2}} \pi^{2}\left(x^{2}, X^{1}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, x^{2}, X^{1}(\omega)\right\}
\end{aligned}
$$

Firm $i$ 's strategy, $X^{i}(\omega)$, is arg max of Bellman equation

$$
\begin{aligned}
& X^{1}(\omega)=\arg \max _{x^{1}} \pi^{1}\left(x^{1}, X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, x^{1}, X^{2}(\omega)\right\} \\
& X^{2}(\omega)=\arg \max _{x^{2}} \pi^{2}\left(x^{2}, X^{1}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, x^{2}, X^{1}(\omega)\right\}
\end{aligned}
$$

## Markov-Perfect Equilibrium

Markov-perfect ("feedback") equilibrium $\left(V^{1}(\omega), x^{1}(\omega), V^{2}(\omega), x^{2}(\omega)\right)$
is a solution to the collection of Bellman equations and strategy equations

Existence: Few applications have existence theorem for pure strategy equilibria Doraszelski and Satterthwaite (2003)

Multiplicity: A common problem; here we aim to find just one
Judd and Schmedders (2005)

## Standard Gauss-Seidel Method

Initialize: Order states $\omega \in \Omega$ and make initial guesses $V^{i}(\omega)$ and $X^{i}(\omega)$
Iterate: Make many passes through $\Omega$, updating values and strategies

$$
\begin{array}{ll}
X^{1}(\omega) \leftarrow \arg \max _{x^{1}} & \pi^{1}\left(x^{1}, X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, x^{1}, X^{2}(\omega)\right\} \\
V^{1}(\omega) \leftarrow \max _{x^{1}} & \pi^{1}\left(x^{1}, X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, x^{1}, X^{2}(\omega)\right\} \\
X^{2}(\omega) \leftarrow \arg \max _{x^{2}} & \pi^{2}\left(x^{2}, X^{1}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, x^{2}, X^{1}(\omega)\right\} \\
V^{2}(\omega) \leftarrow \max _{x^{2}} & \pi^{2}\left(x^{2}, X^{1}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, x^{2}, X^{1}(\omega)\right\}
\end{array}
$$

Basically a best-reply approach
Better than Gauss-Jacobi - a.k.a. value function iteration - which does not update $V^{i}(\omega)$ and $X^{i}(\omega)$ until next iterates are computed at all $\omega$

## Newton Method for Discrete-Time Game

Construct system of equations

- One equation for each value function in each state $\omega$

$$
\begin{aligned}
& V^{1}(\omega)=\pi^{1}\left(X^{1}(\omega), X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, X^{1}(\omega), X^{2}(\omega)\right\} \\
& V^{2}(\omega)=\pi^{2}\left(X^{1}(\omega), X^{2}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, X^{1}(\omega), X^{2}(\omega)\right\}
\end{aligned}
$$

- First-order conditions of optimality of firms' decisions in each state $\omega$

$$
\begin{aligned}
\frac{\partial}{\partial x^{1}}\left(\pi^{1}\left(X^{1}(\omega), X^{2}(\omega), \omega\right)+\beta E\left\{V^{1}\left(\omega^{\prime}\right) \mid \omega, X^{1}(\omega), X^{2}(\omega)\right\}\right) & \leq 0 \\
X^{1}(\omega) & \geq 0 \\
\frac{\partial}{\partial x^{2}}\left(\pi^{2}\left(X^{1}(\omega), X^{2}(\omega), \omega\right)+\beta E\left\{V^{2}\left(\omega^{\prime}\right) \mid \omega, X^{1}(\omega), X^{2}(\omega)\right\}\right) & \leq 0 \\
X^{2}(\omega) & \geq 0
\end{aligned}
$$

## Technical Issues

Large system of nonlinear equations and inequalities

Presence of complementarity conditions

Size of the Jacobian

## Two Firms: Cournot Competition

Two firms produce the same good
In each period firms play a Cournot game and produce quantities $q_{1}, q_{2}$
Total quantity $q=q_{1}+q_{2}$
Inverse demand function $P(q)=A-\phi q$
Firms' cost functions $C_{i}\left(c_{i}, q_{i}\right)=c_{i} q_{i}^{2}$
Technology of firm $i$ given by $c_{i}$
Profits $\pi_{i}$ for firm $i$

$$
\begin{aligned}
& \pi_{1}\left(q_{1}, q_{2} ; c_{1}\right)=q_{1} P\left(q_{1}+q_{2}\right)-c_{1} q_{1}^{2} \\
& \pi_{2}\left(q_{1}, q_{2} ; c_{2}\right)=q_{2} P\left(q_{1}+q_{2}\right)-c_{2} q_{2}^{2}
\end{aligned}
$$

## Static Nash Equilibrium

Static Nash equilibrium can be solved in closed-form

$$
\begin{aligned}
& q_{1}^{N}\left(c_{1}, c_{2}\right)=A \frac{2 c_{2}+\phi}{4 c_{1} c_{2}+4\left(c_{1}+c_{2}\right) \phi+3 \phi^{2}} \\
& q_{2}^{N}\left(c_{1}, c_{2}\right)=A \frac{2 c_{1}+\phi}{4 c_{1} c_{2}+4\left(c_{1}+c_{2}\right) \phi+3 \phi^{2}}
\end{aligned}
$$

Cournot equilibrium profits

$$
\begin{aligned}
\pi_{1}^{N}\left(c_{1}, c_{2}\right) & =\frac{A^{2}\left(c_{1}+\phi\right)\left(2 c_{2}+\phi\right)^{2}}{\left(4 c_{1} c_{2}+4\left(c_{1}+c_{2}\right) \phi+3 \phi^{2}\right)^{2}} \\
\pi_{2}^{N}\left(c_{1}, c_{2}\right) & =\frac{A^{2}\left(c_{2}+\phi\right)\left(2 c_{1}+\phi\right)^{2}}{\left(4 c_{1} c_{2}+4\left(c_{1}+c_{2}\right) \phi+3 \phi^{2}\right)^{2}}
\end{aligned}
$$

## Dynamic Model

Firm $i$ can affect production cost $c_{i}$ through investment
For simplicity: $c_{i}=\frac{1}{M_{i}}$ where $M_{i}$ is the number of machines of firm $i$
$M_{i}$ depends on investment effort and depreciation
Increase in $M_{i}$ through investment, decrease in $M_{i}$ through depreciation
Probability of depreciation shock $\delta$

Cost of investment effort $u_{i}$ is $C_{i}\left(u_{i}\right)=\gamma_{i} u_{i}+\eta_{i}\left(u_{i}\right)^{2}$
Observe $C_{i}^{\prime}(0)=\gamma_{i}$

Distinguish production cost $c_{i}=\frac{1}{M_{i}}$ and investment $\operatorname{cost} C_{i}\left(u_{i}\right)$

## Stochastic Transition Process

Number of machines $M_{i} \in\{1,2, \ldots, N\}$
Popular specification of transition probabilities for $2 \leq M_{i} \leq N-1$

$$
\operatorname{Pr}^{i}\left(M_{i}^{+} \mid M_{i}, u_{i}\right)=\left\{\begin{array}{cl}
\frac{(1-\delta) \alpha u_{i}}{1+\alpha u_{i}} & \xi^{i}=M_{i}+1 \\
\frac{1-\delta+\delta \alpha u_{i}}{1+\alpha u_{i}} & \xi^{i}=M_{i} \\
\frac{\delta}{1+\alpha u_{i}} & \xi^{i}=M_{i}-1
\end{array}\right.
$$

State-to-state transition probabilities

$$
\operatorname{Pr}\left(\left(M_{1}^{+}, M_{2}^{+}\right) \mid\left(M_{1}, M_{2}\right),\left(u_{1}, u_{2}\right)\right)=\operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, u_{1}\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, u_{2}\right)
$$

## Complete Dynamic Game

State of the economy is $\left(M_{1}, M_{2}\right)$ at the beginning of period
Production technologies of firms $\left(c_{1}, c_{2}\right)=\left(\frac{1}{M_{1}}, \frac{1}{M_{2}}\right)$
Cournot outcome on product market with period profits $\left(\pi_{1}^{N}, \pi_{2}^{N}\right)$
Firms' investment in technology $\left(u_{1}, u_{2}\right)$ incurring costs $\left(C_{1}\left(u_{1}\right), C_{2}\left(u_{2}\right)\right)$
Stochastic transition to new states $\left(M_{1}^{+}, M_{2}^{+}\right)$for next period

Infinite-horizon model
Firms have discount factor $\beta$
Firms maximize expected discounted sum of per-period profits

## Optimality Conditions

Separation between static Cournot game and dynamic investment decisions
Optimal investment effort $U_{1}\left(M_{1}, M_{2}\right)$ satisfies

$$
\begin{aligned}
& V_{1}\left(M_{1}, M_{2}\right)=\left(\pi_{1}^{N}\left(M_{1}, M_{2}\right)-C_{1}\left(U_{1}\left(M_{1}, M_{2}\right)\right)\right) \\
& \quad+\beta \sum_{M_{1}^{+}} \sum_{M_{2}^{+}} \operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, U_{1}\left(M_{1}, M_{2}\right)\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, U_{2}\left(M_{1}, M_{2}\right)\right) V_{1}\left(M_{1}^{+}, M_{2}^{+}\right)
\end{aligned}
$$

If $U_{1}\left(M_{1}, M_{2}\right)>0$ then

$$
\begin{aligned}
0 & =-\frac{\partial}{\partial u_{1}} C_{1}\left(U_{1}\left(M_{1}, M_{2}\right)\right) \\
& +\beta \sum_{M_{1}^{+}} \sum_{M_{2}^{+}} \frac{\partial}{\partial u_{1}} \operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, U_{1}\left(M_{1}, M_{2}\right)\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, U_{2}\left(M_{1}, M_{2}\right)\right) V_{1}\left(M_{1}^{+}, M_{2}^{+}\right)
\end{aligned}
$$

If $U_{1}\left(M_{1}, M_{2}\right)=0$ then

$$
\begin{aligned}
0 & \geq-\frac{\partial}{\partial u_{1}} C_{1}\left(U_{1}\left(M_{1}, M_{2}\right)\right) \\
& +\beta \sum_{M_{1}^{+}} \sum_{M_{2}^{+}} \frac{\partial}{\partial u_{1}} \operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, U_{1}\left(M_{1}, M_{2}\right)\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, U_{2}\left(M_{1}, M_{2}\right)\right) V_{1}\left(M_{1}^{+}, M_{2}^{+}\right)
\end{aligned}
$$

## Solutions

Recall cost of investment effort $C_{1}\left(u_{1}\right)=\gamma_{1} u_{1}+\eta_{1}\left(u_{1}\right)^{2}$
If $\gamma_{1}=0$ then interior solution $u_{1}>0$ and no complementarity conditions necessary
If $\gamma_{1}>0$ then boundary solution $u_{1}=0$ possible

Four equations for each state $\left(M_{1}, M_{2}\right)$, so $4 \times N^{2}$ equations

Running times in seconds (using the PATH solver)

| $\gamma_{1}$ | $\gamma_{2}$ | $N=20$ | $N=50$ | $N=80$ | $N=100$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 0 | 0.56 | 11.2 | 72 | 146 |
| 1 | 1 | 0.57 | 12.5 | 59 | 192 |
| 1 | 2 | 0.62 | 12.8 | 98 | 182 |

## More Interesting Models

Cournot stage game was solved in closed-form
No analytical solution for Cournot quantity $q_{i}$ for more general functions
Replace

$$
\begin{aligned}
& V_{1}\left(M_{1}, M_{2}\right)=\arg \max _{u_{1}}\left(\pi_{1}^{N}\left(M_{1}, M_{2}\right)-C_{1}\left(u_{1}\right)\right) \\
& \quad+\beta \sum_{M_{1}^{+}} \sum_{M_{2}^{+}} \operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, u_{1}\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, U_{2}\left(M_{1}, M_{2}\right)\right) V_{1}\left(M_{1}^{+}, M_{2}^{+}\right)
\end{aligned}
$$

by

$$
\begin{aligned}
& V_{1}\left(M_{1}, M_{2}\right)=\arg \max _{u_{1}, q_{1}}\left(\pi_{1}\left(q_{1}, Q_{2}\left(M_{1}, M_{2}\right) ; M_{1}\right)-C_{1}\left(u_{1}\right)\right) \\
& \quad+\beta \sum_{M_{1}^{+}} \sum_{M_{2}^{+}} \operatorname{Pr}^{1}\left(M_{1}^{+} \mid M_{1}, u_{1}\right) \cdot \operatorname{Pr}^{2}\left(M_{2}^{+} \mid M_{2}, U_{2}\left(M_{1}, M_{2}\right)\right) V_{1}\left(M_{1}^{+}, M_{2}^{+}\right)
\end{aligned}
$$

where

$$
\pi_{1}\left(q_{1}, q_{2} ; M_{1}\right)=q_{1} P\left(q_{1}+q_{2}\right)-\frac{1}{M_{1}} q_{1}^{2}
$$

## More Equations

Additional optimality conditions

If $Q_{1}\left(M_{1}, M_{2}\right)>0$ then

$$
\frac{\partial}{\partial q_{1}} \pi_{1}\left(Q_{1}\left(M_{1}, M_{2}\right), Q_{2}\left(M_{1}, M_{2}\right) ; M_{1}\right)=0
$$

If $Q_{1}\left(M_{1}, M_{2}\right)=0$ then

$$
\frac{\partial}{\partial q_{1}} \pi_{1}\left(Q_{1}\left(M_{1}, M_{2}\right), Q_{2}\left(M_{1}, M_{2}\right) ; M_{1}\right) \leq 0
$$

Additional complementarity conditions

## Solving More Equations

Six equations for each state $\left(M_{1}, M_{2}\right)$, so $6 \times N^{2}$ equations

Production quantities are always positive (complementarity conditions not needed)

Running times in seconds

| $\gamma_{1}$ | $\gamma_{2}$ | $N=20$ | $N=50$ | $N=80$ | $N=100$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 0 | 0.65 | 13.9 | 62 | 128 |
| 1 | 1 | 0.65 | 14.4 | 112 | 287 |
| 1 | 2 | 0.70 | 15.3 | 86 | 234 |

No significant difference to smaller systems with explicit profit functions

## Summary

Stochastic dynamic discrete-time games with thousands of states

Explicit solution for the static Nash equilibrium unnecessary

Multi-dimensional controls

Complementarity conditions

Corner solutions

## Next Steps

More general cost functions

Multi-dimensional state vectors per player

More players

More general transitions (jump more than one unit per state)

Specialized version of PATH: better scaling and linear algebra routines

