# EABCN Training School: Monetary-Fiscal Policy Interactions 

Lecture 6. Foresight: Theory and Econometrics

## Eric M. Leeper

Indiana University
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## The Messages

- Draws on Leeper, Walker, Yang $(2008,2009)$
- It's about the problems that arise whenever agents' information sets do not align with the econometrician's information set
- Implications
- standard econometric tools will fail
- inferences can be very misleading
- no quick fix to the problem
- Need to model information flows in a serious way
- This lecture will use fiscal foresight as the example
- Applies to rapidly growing literature on technology foresight


## Fiscal Foresight: The Problem

- Legislative \& implementation lags $\Rightarrow$ agents know changes in future tax rates before they are effective
- Agents act on the information before fiscal variables move
- Hard to build agents' information into econometric work
- Cannot extract "news" from current \& past fiscal variables


## Implications of Fiscal Foresight

- Agents' \& econometrician's information sets misaligned
- Conventional econometric methods can fail to identify "news" correctly
- Fiscal foresight can create a non-invertible moving average in equilibrium data (first shown by Yang)
- Usual econometric tools can yield false inferences
- impulse response functions, variance decompositions
- tests of cross-equation restrictions
- tests of present-value relations
- All identifications convoluted: fiscal \& non-fiscal
- dynamics wrong
- shocks confounded


## A Bit of Formalism

- $\varepsilon_{t}$ : vector of exogenous shocks agents observe
- $\varepsilon_{\tau, t}$ : tax component
- $\Omega_{t}=\operatorname{span}\left\{\varepsilon_{t}, \varepsilon_{t-1}, \ldots\right\}$ agents' information set
- $\varepsilon_{t}^{*}$ : vector of exogenous shocks econometrician identifies
- $\varepsilon_{\tau, t}^{*}$ : tax component
- $\Omega_{t}^{*}=\operatorname{span}\left\{\varepsilon_{t}^{*}, \varepsilon_{t-1}^{*}, \ldots\right\}$ : econometrician's information set
- Fiscal foresight $\Rightarrow \Omega_{t}^{*}$ strictly smaller than $\Omega_{t}$


## No Consensus in Empirical Work

An anticipated tax cut

- has little or no effect [Poterba-Summers, Blanchard-Perotti, Romer-Romer]
- is expansionary in the short run [Mountford-Uhlig]
- is strongly contractionary in the short run [Mertens-Ravn, Branson-Fraga-Johnson, House-Shapiro]

Sources of the diverse results

- invalid instruments for fiscal foresight
- no modeling of fiscal behavior that gives rise to foresight (information flows)
- non-invertibility not directly confronted


## Anecdotal Evidence of Fiscal Foresight

- Large public finance literature
- Branson, Fraga, \& Johnson: Economic Recovery Tax Act of 1981 "purely anticipatory recession"
- House \& Shapiro: jobless recovery of 2001-tax phase-ins induced production delays
- Ramey: War dummies predict defense spending


## Capital Gains in Anticipation of TRA86

| Year | Long-term | Total |
| :---: | :---: | :---: |
| 1984 | 134.1 | 135.0 |
| 1985 | 164.9 | 167.0 |
| 1986 | $\mathbf{3 1 5 . 7}$ | $\mathbf{3 2 2 . 2}$ |
| 1987 | - | 137.4 |
| 1988 | - | 153.8 |
| 1989 | - | 145.6 |
| 1990 | - | 113.2 |

Capital Gains Realizations in Billions. Source: Auerbach \& Slemrod (1997)

## Funds Rate in Anticipation of TRA86



## Daily Federal Funds Rate

## Funds Rate in Anticipation of TRA86

"Despite repeated reserve injections, reserve needs persisted...The Treasury balance was enlarged beyond normal levels by heavy sales of nonmarketable debt to tax-exempt authorities which were engaged in large sales in the market to get ahead of proposed restrictive tax legislation."
-FRBNY Quarterly Review Spring 1986

## Simple Illustrative Model

- Log preferences
- Inelastic labor supply
- Complete depreciation of capital
- Proportional tax levied against income $\left[T_{t}=\tau_{t} Y_{t}\right]$

Equilibrium conditions

$$
\begin{gathered}
\frac{1}{C_{t}}=\alpha \beta \mathbf{E}_{\mathrm{t}}\left(\mathbf{1}-\tau_{\mathrm{t}+1}\right) \frac{1}{C_{t+1}} \frac{Y_{t+1}}{K_{t}} \\
C_{t}+K_{t}=Y_{t}=A_{t} K_{t-1}^{\alpha}
\end{gathered}
$$

## Solving the Model

- Log linearize to get a second-order difference equation in $k$

$$
\begin{aligned}
& E_{t} k_{t+1}-\frac{1+\alpha^{2} \beta(1-\tau)}{\alpha \beta(1-\tau)} k_{t}+\frac{1}{\beta(1-\tau)} k_{t-1} \\
= & \frac{1-\alpha \beta(1-\tau)}{\alpha \beta(1-\tau)}\left(\frac{\tau}{1-\tau}\right) E_{t} \hat{\tau}_{t+1}-\frac{1}{\alpha \beta(1-\tau)} a_{t}
\end{aligned}
$$

or

$$
E_{t} k_{t+1}-\gamma_{0} k_{t}+\gamma_{1} k_{t-1}=\nu_{0} \varepsilon_{A, t}+\nu_{1} E_{t} \hat{\tau}_{t+1}
$$

with

$$
\begin{aligned}
& \gamma_{0}=\frac{1+\alpha^{2} \beta(1-\tau)}{\alpha \beta(1-\tau)}>0, \quad \gamma_{1}=\frac{1}{\beta(1-\tau)}>0 \\
& \nu_{0}=-\frac{1}{\alpha \beta(1-\tau)}<0, \quad \nu_{1}=\frac{1-\alpha \beta(1-\tau)}{\alpha \beta(1-\tau)}\left(\frac{\tau}{1-\tau}\right)>0
\end{aligned}
$$

## Solving the Model

- Solution satisfies saddlepath property; write the difference equation as

$$
\left(B^{-2}-\gamma_{0} B^{-1}+\gamma_{1}\right) E_{t} k_{t-1}=E_{t} z_{t}
$$

where $B^{-j} E_{t-1} k_{t}=E_{t-1} k_{t+j}$ for integer $j$ and

$$
z_{t} \equiv \nu_{0} \varepsilon_{A, t}+\nu_{1} E_{t} \hat{\tau}_{t+1}
$$

- Factor the quadratic as

$$
\left(\lambda_{1}-B^{-1}\right)\left(\lambda_{2}-B^{-1}\right) E_{t} k_{t-1}=E_{t} z_{t}
$$

so that $\gamma_{1}=\lambda_{1} \lambda_{2}$ and $\gamma_{0}=\lambda_{1}+\lambda_{2}$. Note that $\lambda_{1}>0$ and $\lambda_{2}>0$

- Select $\lambda_{1}<1$ and $\lambda_{2}=\left[\beta(1-\tau) \lambda_{1}\right]^{-1}>1$
- Operate on both sides of the equation with $\left(\lambda_{2}-B^{-1}\right)^{-1}$

$$
\left(\lambda_{1}-B^{-1}\right) E_{t} k_{t-1}=\left(\lambda_{2}-B^{-1}\right)^{-1} E_{t} z_{t}
$$

## Solving the Model

- Now

$$
\frac{1}{\lambda_{2}-B^{-1}}=\frac{1}{\lambda_{2}} \frac{1}{1-\left(1 / \lambda_{2}\right) B^{-1}}
$$

We shall use the facts that $\lambda_{2}^{-1}=\beta(1-\tau) \lambda_{1}$ and $\left[1-\left(1 / \lambda_{2}\right) B^{-1}\right]^{-1}=\sum_{j=0}^{\infty}\left(\lambda_{2} B\right)^{-j}$ to yield

$$
k_{t}=\lambda_{1} k_{t-1}-\beta(1-\tau) \lambda_{1} \sum_{i=0}^{\infty}\left[\beta(1-\tau) \lambda_{1}\right]^{i} E_{t} z_{t+i}
$$

- It turns out that $\lambda_{1}=\alpha<1 \& \lambda_{2}=[\alpha \beta(1-\tau)]^{-1}>1$
- The solution for $k_{t}$ is a function of the state at $t: k_{t-1}$ and current and expected exogenous disturbances known at $t$


## Model Solution

Equilibrium capital accumulation obeys

$$
k_{t}=\alpha k_{t-1}+a_{t}-(1-\theta)\left(\frac{\tau}{1-\tau}\right) \sum_{\mathrm{i}=0}^{\infty} \theta^{\mathrm{i}} \mathrm{E}_{\mathrm{t}} \hat{\tau}_{\tau, \mathrm{t}+1+\mathrm{i}}
$$

where $\theta=\alpha \beta(1-\tau)<1$ and $a_{t}$ is exogenous technology

- $\theta$ plays central role in analysis
- Agent uses $\theta$ to discount tax rates in usual way
- How does agent discount tax news?


## Fiscal Foresight

- Need to specify information flows
- Start with simple flow
- Tax news arrives $q$ periods before tax rates change

$$
\hat{\tau}_{t}=\varepsilon_{\tau, t-q}
$$

- Technology: i.i.d. so $a_{t}=\varepsilon_{A, t}$
- Agent's information set at $t$ consists of variables dated $t$ and earlier, including i.i.d. exogenous shocks

$$
\Omega_{t}=\left\{\varepsilon_{A, t-j}, \varepsilon_{\tau, t-j}\right\}_{j=0}^{\infty}
$$

- Agent at $t$ has (perfect) knowledge of $\left\{\hat{\tau}_{t+q}, \hat{\tau}_{t+q-1}, \ldots\right\}$


## Solution: Various Degrees of Foresight

$q=0$ implies:

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}
$$

$q=1$ implies:

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}-(1-\theta)\left(\frac{\tau}{1-\tau}\right) \varepsilon_{\tau, t}
$$

$q=2$ implies:

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}-(1-\theta)\left(\frac{\tau}{1-\tau}\right)\left\{\varepsilon_{\tau, t-1}+\theta \varepsilon_{\tau, t}\right\}
$$

$q=3$ implies:

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}-(1-\theta)\left(\frac{\tau}{1-\tau}\right)\left\{\varepsilon_{\tau, t-2}+\theta \varepsilon_{\tau, t-1}+\theta^{2} \varepsilon_{\tau, t}\right\}
$$

## DISCOUNTING

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}-(1-\theta)\left(\frac{\tau}{1-\tau}\right)\left\{\varepsilon_{\tau, t-1}+\theta \varepsilon_{\tau, t}\right\}
$$

- More recent news is discounted (by $\theta$ ) relative to more distant news. Why?
- $\varepsilon_{\tau, t-1}$ affects $\hat{\tau}_{t+1}$
- $\varepsilon_{\tau, t}$ affects $\hat{\tau}_{t+2}$
- news that affects taxes farther into the future is discounted heaviest
- Foresight introduces moving-average terms into equilibrium
- Dynamic optimization implies more recent news more heavily discounted
- Seems perverse and creates econometric problems


## VARs and Foresight

- Linear environment \& Gaussian random variables $\Rightarrow$ projections equivalent to conditional expectations
- VAR is projection $P\left[\mathbf{x}_{t} \mid \mathbf{x}_{t-1}, \mathbf{x}_{t-2}, \ldots\right]$
- If $\mathbf{x}_{t-1}, \mathbf{x}_{t-2}, \ldots$ reveals agents' information set, then econometrician captures dynamics of economy
- Foresight implies $\mathbf{x}_{t-1}, \mathbf{x}_{t-2}, \ldots$ will not capture information set of agent in "typical" VAR
- Agent's information set at $t: \Omega_{t}=\left\{\varepsilon_{A, t-j}, \varepsilon_{\tau, t-j}\right\}_{j=0}^{\infty}$

Econometrician's information set at $t: \Omega_{t}^{*}=\left\{k_{t-j}, z_{t-j}\right\}_{j=0}^{\infty}$

## Aside: Representation Theory

- Univariate example: ARMA

$$
\begin{equation*}
x_{t}=\left[\frac{L-\theta}{1-\rho L}\right] \varepsilon_{t}, \quad|\rho| \in(0,1),|\theta| \in(0,1) \tag{1}
\end{equation*}
$$

- Given $|\theta| \in(0,1),(1)$ is not invertible $\Rightarrow$ linear space spanned by $\left\{x_{t-j}\right\}_{j=0}^{\infty}$ is not equal to the linear space spanned by $\left\{\varepsilon_{t-j}\right\}_{j=0}^{\infty}$
- To find space spanned by $\left\{x_{t-j}\right\}_{j=0}^{\infty}$, need to factor ARMA

$$
\begin{gather*}
x_{t}=\left[\frac{L-\theta}{1-\rho L}\right]\left[\frac{1-\theta L}{L-\theta}\right]\left[\frac{L-\theta}{1-\theta L}\right] \varepsilon_{t} \\
x_{t}=\left[\frac{1-\theta L}{1-\rho L}\right] e_{t}  \tag{2}\\
e_{t}=\left[\frac{L-\theta}{1-\theta L}\right] \varepsilon_{t} \tag{3}
\end{gather*}
$$

- (2) is invertible: current \& past $x_{t}$ span same space as current \& past $e_{t}$ (but not $\varepsilon_{t}$ )


## Aside: Representation Theory

- This points out that the agent's and the econometrician's information sets are different
- Agent observes $\left\{\varepsilon_{t}\right\}$
- Econometrician observes $\left\{x_{t}\right\}$
- Agent's information set larger than econometrician's
- $\varepsilon_{t}$ called "non-fundamental" shocks because they produce a non-invertible representation
- $e_{t}$ called "fundamental" shocks because they are associated with invertible (Wold) representation


## Aside: Representation Theory

- $(1-\theta L) /(L-\theta)$ is a Blaschke factor
- From (2), $x_{t}=\left[\frac{1-\theta L}{1-\rho L}\right] e_{t}$
- current \& past $\varepsilon_{t}$ sufficient for $e_{t}$
- but inverse of Blaschke factor does not possess a valid expansion inside the unit circle in $L$ due to the pole at

$$
L=|\theta|
$$

- hence, current \& past $e_{t}$ do not reveal $\varepsilon_{t}$
- Setting $F=L^{-1}$, Blaschke factor has valid inverse in the forward operator $F$

$$
\left[\frac{F-\theta}{1-\theta F}\right] e_{t}=\varepsilon_{t}, \quad \varepsilon_{t}=\left(L^{-1}-\theta\right) \sum_{j=0}^{\infty} \theta^{j} e_{t+j}
$$

- hence $\varepsilon_{t}$ carries information about future $e$ 's (and $x$ 's)


## Aside: Representation Theory

- A famous example

$$
y_{t}=w_{t}+2 w_{t-1}, \quad w_{t} \sim \operatorname{iid} N(0,1)
$$

- Define VAR innovations at $a_{t}=y_{t}-\hat{E}\left(y_{t} \mid y_{t-1}, y_{t-2}, \ldots\right)$
- MA representation of $y_{t}$ is

$$
y_{t}=2\left(a_{t}+(1 / 2) a_{t-1}\right), \quad a_{t} \sim i i d N(0,1)
$$

- IRF of first model, ( $1,2,0,0, \ldots$ ) different from VAR, $(2,1,0,0, \ldots)$ : Why?
- $a_{t}$ belongs to linear space spanned by $y_{t}$ :

$$
a_{t}=\frac{1}{2} \sum_{j=0}^{\infty}\left(-\frac{1}{2}\right)^{j} y_{t-j}
$$

- and $w_{t}$ belongs to the linear space spanned by future $y_{t}$ :

$$
w_{t}=\frac{1}{2} \sum_{j=0}^{\infty}\left(-\frac{1}{2}\right)^{j} y_{t+j+1}
$$

## Foresight \& Non-Invertibility

- Econometrician's conditioning set: $\left\{k_{t-j}, a_{t-j}\right\}_{j=0}^{\infty}$
- Will drop $a_{t}$ from equations since econometrician knows it
- Best case scenario
- Solution with 2-period foresight

$$
(1-\alpha L) k_{t}=-\kappa(L+\theta) \varepsilon_{\tau, t}
$$

- Does $\left\{k_{t-j}\right\}_{j=0}^{\infty} \equiv\left\{\varepsilon_{\tau, t-j}\right\}_{j=0}^{\infty}$ ?
- Invertibility requires $|\theta|>1$ : yields conv. seq. in past $k$

$$
\left[\frac{1-\alpha L}{1+\boldsymbol{\theta}^{-1} \mathbf{L}}\right] k_{t}=-\kappa \theta \varepsilon_{\tau, t}
$$

But $\theta<1$, so not invertible in current and past capital

- Is invertible in current and future capital

$$
\begin{aligned}
k_{t}= & \left(\alpha^{-1}+\theta\right) k_{t+1}-\theta\left(\alpha^{-1}+\theta\right) k_{t+2}+ \\
& \theta^{2}\left(\alpha^{-1}+\theta\right) k_{t+3}-\cdots+\kappa \varepsilon_{\tau, t}
\end{aligned}
$$

## Econometrician's Estimates: I

- So $\left\{k_{t-j}\right\}_{j=0}^{\infty} \neq\left\{\varepsilon_{\tau, t-j}\right\}_{j=0}^{\infty}$
- Need to find the econometrician's information set: $\left\{k_{t-j}\right\}_{j=0}^{\infty} \equiv$ ???
- Wold representation for capital

$$
\begin{aligned}
(1-\alpha L) k_{t} & =\underbrace{-\kappa(L+\theta)\left[\frac{1+\theta L}{L+\theta}\right]}_{-\kappa(\mathbf{1}+\theta \mathbf{L})} \underbrace{\left[\frac{L+\theta}{1+\theta L}\right] \varepsilon_{\tau, t}}_{\varepsilon_{\tau, \mathrm{t}}^{*}} \\
& =-(1-\theta)\left(\frac{\tau}{1-\tau}\right)\left\{\theta \varepsilon_{\tau, t-1}^{*}+\varepsilon_{\tau, t}^{*}\right\}
\end{aligned}
$$

## Econometrician's Discounting I

- Econometrician does not discount news same way as agent
- Econometrician recovers current and past $\varepsilon_{\tau}^{*}$ not $\varepsilon_{\tau}$
- Econometrician's innovations are "old news"

$$
\begin{gathered}
\varepsilon_{\tau, t}^{*}=\theta \varepsilon_{\tau, t}+\left(1-\theta^{2}\right) \varepsilon_{\tau, t-1}-\theta\left(1-\theta^{2}\right) \varepsilon_{\tau, t-2}+ \\
\theta^{2}\left(1-\theta^{2}\right) \varepsilon_{\tau, t-3}+\cdots
\end{gathered}
$$

- Econometrician discounts innovations incorrectly because information set lags agents'
- econometrician: $k_{t}$ depends on $\theta \varepsilon_{\tau, t-1}^{*}+\varepsilon_{\tau, t}^{*}$
- agents: $k_{t}$ depends on $\varepsilon_{\tau, t-1}+\theta \varepsilon_{\tau, t}$


## Impulse Response Functions: I



## Impulse Response Functions: I



## Econometrician's Estimates: II

- Make more plausible assumption that econometrician does not observe technology: $\left\{\hat{\tau}_{t-j}, k_{t-j}\right\}_{j=0}^{\infty}$
- Does $\left\{\hat{\tau}_{t-j}, k_{t-j}\right\}_{j=0}^{\infty} \equiv\left\{\varepsilon_{\tau, t-j}, \varepsilon_{A, t-j}\right\}_{j=0}^{\infty}$ ?
- No: Econometrician's shocks convolute agents' news

$$
\begin{gathered}
\varepsilon_{\tau, t}^{*}=a_{1} \varepsilon_{\tau, t-1}+a_{2} \varepsilon_{\tau, t-2}+a_{3} \varepsilon_{A, t-1}+a_{4} \varepsilon_{A, t-2} \\
\varepsilon_{A, t}^{*}=b_{1} \varepsilon_{\tau, t}+b_{2} \varepsilon_{\tau, t-1}+b_{3} \varepsilon_{A, t}+b_{4} \varepsilon_{A, t-1}
\end{gathered}
$$

$a$ 's and $b$ 's are functions of model parameters

- Econometrician gets effects of both taxes and technology wrong
- Conclude taxes don't matter; everything driven by technology


## Impulse Response Functions: II



## Impulse Response Functions: II



## Impulse Response Functions: III

- Have shown that depending on econometrician's information set, may infer that higher expected taxes
- are contractionary (Mountford \& Uhlig)
- have little effect (Poterba, Blanchard \& Perotti, Romer \& Romer)
- Now show may infer higher expected taxes are expansionary (Mertens \& Ravn, House \& Shapiro)
- Consider effects of variations in $\sigma_{a} / \sigma_{\tau}$ (relative volatility of technology and taxes)
- alters the signal-extraction problem econometrician faces
- as $\sigma_{a} / \sigma_{\tau} \rightarrow 0$, problem gets worse
- infer higher expected taxes are increasingly expansionary


## Impulse Response Functions: III



Responses of $k$ to Higher Expected Taxes

## Impulse Response Functions: III



Responses of $k$ to Higher Expected Taxes

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## Impulse Response Functions: III



Responses of $k$ to Higher Expected Taxes

## Impulse Response Functions: III



Responses of $k$ to Higher Expected Taxes

## ABCD Test for Invertibility

- Consider the system with 2 period foresight whose eqm is

$$
\left[\begin{array}{c}
\tau_{t} \\
k_{t}
\end{array}\right]=\left[\begin{array}{cc}
L^{2} & 0 \\
-\frac{\kappa(L+\theta)}{1-\alpha L} & \frac{1}{1-\alpha L}
\end{array}\right]\left[\begin{array}{c}
\varepsilon_{\tau, t} \\
\varepsilon_{A, t}
\end{array}\right]
$$

- We showed by directly computing the roots of the MA term that this is not invertible
- Fernandez-Villaverde, Rubio-Ramirez, Sargent, Watson propose a simple test of invertibility for a system written in state-space form
- This test will give identical results as checking roots of MA

$$
\begin{aligned}
& \text { ABCD Test: State-Space Form } \\
& \begin{aligned}
{\left[\begin{array}{c}
\tau_{t+1} \\
k_{t+1} \\
\varepsilon_{\tau, t+1} \\
\varepsilon_{\tau, t}
\end{array}\right] } & =\left[\begin{array}{cccc}
0 & 0 & 0 & 1 \\
0 & \alpha & -\theta & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{c}
\tau_{t} \\
k_{t} \\
\varepsilon_{\tau, t} \\
\varepsilon_{\tau, t-1}
\end{array}\right]+\left[\begin{array}{cc}
0 & 0 \\
-\kappa \theta & 1 \\
1 & 0 \\
0 & 0
\end{array}\right]\left[\begin{array}{l}
\varepsilon_{\tau, t+1} \\
\varepsilon_{A, t+1}
\end{array}\right] \\
x_{t+1} & =A x_{t}+B w_{t+1}
\end{aligned} \\
& {\left[\begin{array}{c}
\tau_{t+1} \\
k_{t+1}
\end{array}\right]=\left[\begin{array}{cccc}
0 & 0 & 0 & 1 \\
0 & \alpha & -\theta & 0
\end{array}\right]\left[\begin{array}{c}
\tau_{t} \\
k_{t} \\
\varepsilon_{\tau, t} \\
\varepsilon_{\tau, t-1}
\end{array}\right]+\left[\begin{array}{cc}
0 & 0 \\
-\kappa \theta & 1
\end{array}\right]\left[\begin{array}{c}
\varepsilon_{\tau, t+1} \\
\varepsilon_{A, t+1}
\end{array}\right]} \\
& y_{t+1}=C x_{t}+D w_{t+1}
\end{aligned}
$$

- If $D^{-1}$ exists, then the system is invertible if and only if all the eigenvalues of $A-B D^{-1} C$ are inside the unit circle
- With foresight and no unanticipated part to taxes, $D \mathrm{~s}$ singular and ABCD test cannot be applied
- Are straightforward ways to make $D$ non-singular
- add an unanticipated contemporaneous shock to tax rule: $e_{t}^{u}$
- allow automatic stabilizers: $\tau_{t}=\phi y_{t}+\varepsilon_{\tau, t-q}$


## INTERMISSION

- As vary econometrician's information set
- by adding or subtracting data
- by changing signal-extraction problem
- Can obtain any inference about effects of news about higher future taxes
- Connects to findings in empirical literature


## Anticipated Tax Hike: Contractionary



## Anticipated Tax Hike: No Effect



## Anticipated Tax Hike: Expansionary



## Robustness

1. Do results from simple model carry over to more complex settings?

- extend model specification-a "serious" model
- generalize information flows

2. How do model elements alter the effects of foresight?

- internal propagation mechanisms alter econometric errors in important ways

3. How sensitive are results to alternative assumptions about information flows?

- relax rigid assumptions above and in "news" literature


## Information Flows-Intuition

- Tax process

$$
\begin{aligned}
& \hat{\tau}_{t}=\rho_{1} \hat{\tau}_{t-1}+\cdots+\rho_{n} \hat{\tau}_{t-n}+\varepsilon_{\tau, t-2} \\
& \alpha \beta(1-\tau)<\left(1+\rho_{1}\right)^{-1} \Rightarrow \text { non-invertibility. }
\end{aligned}
$$

- Tax process

$$
\hat{\tau}_{t}=\psi \varepsilon_{\tau, t-2}+(1-\psi) \varepsilon_{\tau, t-1} \quad \psi \in(0,1)
$$

- Capital Dynamics

$$
k_{t}=\alpha k_{t-1}+\varepsilon_{A, t}-\left(\frac{(1-\theta) \tau}{1-\tau}\right)\left\{[1-\psi(1-\theta)] \varepsilon_{\tau, t}+\psi \varepsilon_{\tau, t-1}\right\}
$$

- If more recent news receives the heavier discount, $1-\psi(1-\theta)<\psi$, then the equilibrium will be non-invertible.


## A Serious Model

- Includes elastic labor supply, variable utilization rates for capital inputs, durable and non-durable consumption, habit formation in non-durable consumption, investment adjustment costs, deliberation costs for durable, capital and labor taxes goods
- Calibrated to US data 1947:Q1 to 2008:Q2
- Tax Rules: Generalizing information flows

$$
\widehat{\tau}_{t}^{i}=\rho_{i} \widehat{\tau}_{t-1}^{i}+q_{i} \widehat{s}_{t-1}^{B}+\mu_{i} \widehat{Y}_{t}+\sum_{j=0}^{4} \phi_{j} \varepsilon_{i, t-j}
$$

with $\sum_{j} \phi_{j}=1$ for $i=K, L$

- $\phi$ 's are weights that imply trajectories of expected tax rates
- determined by technology of tax choice


## Information Flows and Tax Multipliers

 A class of information flows-the $\phi$ 's in $\sum_{j} \phi_{j} \varepsilon_{i, t-j}$1. Use serious model and randomly draw $\phi$ 's according to:

$$
\phi_{4} \sim 1-\beta(1.5,5) ; \quad \phi_{1}, \phi_{2}, \phi_{3} \sim U[-0.1,0.1]
$$

2. $\phi_{4}$ is left skewed between 0 and 1 with mean of 0.77 and standard deviation of 0.15 following work on tax information flows by Yang (2008)

- very conservative period of foresight
- $\phi$ 's reflect average degree of foresight

3. Need not imply non-fundamental representation
4. Estimate identified VAR and calculate dynamic multipliers

- obtain distribution for degree of foresight
- derive distribution for errors in inference


## Information Flows and Tax Multipliers

- Move beyond $0-1$ treatment of noninvertibility
- Ask if errors of inference are quantitatively important
- Empirical estimates of multipliers from anticipated tax changes are all over the map:

1. Poterba, Blanchard \& Perotti, Romer \& Romer: $\approx 0$
2. Mountford \& Uhlig: strongly negative
3. Mertens \& Ravn: strongly positive

## Information Flows and Tax Multipliers


$K$-Tax Output Multipliers: Actual \& Estimated

## Information Flows and Tax Multipliers



## K-Tax Consumption Multipliers: Actual \& Estimated

## Solutions: The Usual Suspects

- Expand the econometrician's information set
- add financial variables [Sims, Beaudry-Portier]
- add revenue forecasts [Romer-Romer]
- need strong identifying assumptions in either case
- Impose sign restrictions to identify a VAR
- theory supports any response of economic activity to anticipated tax increase
- Estimate a DSGE model
- fine; conditional on getting information flows right [Blanchard-L’Huillier-Lorenzoni, Christiano et al., Schmitt-Grohe/Uribe]


## Broader Implications

- Analysis extends to other areas where information flows emphasized
- News about future technological improvement [Beaudry-Portier; Christiano-llut-Motto-Rostagno; Jaimovich-Rebelo; Schmitt-Grohe/Uribe]
- Foresight about government spending run ups [Ramey-Shapiro, Ramey]
- Inflation-targeting central banks that publish interest rate paths [Laseen-Linde-Svensson]
- Distinction between "authorization" (e.g., 2009) and actual "outlays" (through 2019), especially for government infrastructure spending [Leeper-Walker-Yang]


## Two Empirical Lines of Attack

- Ex-post: estimate conventional VARs
- creative identification of "anticipated taxes"
- Sims, Blanchard-Perotti, Mountford-Uhlig, Yang
- Ex-ante: reject VARs
- identify foresight through narrative method
- Ramey-Shapiro, Ramey, Romer-Romer, Mertens-Ravn
- Both lines seek instruments for foresight (future $k$ )
- We assess the methods
- formalize "narrative" approach
- theory leads to skepticism about the methods


## Ex-Post Approach

- Estimate VAR ignoring foresight and then impose identification restrictions to deal with foresight
- Tend to conclude foresight is second-order [Mountford-Uhlig and Blanchard-Perotti]
- But if foresight not handled properly, variation due to anticipated shocks gets attributed to unanticipated shocks
- consider the following tax rule in the simple model

$$
\hat{\tau}_{t}=\mathrm{e}_{\mathrm{t}}^{\mathrm{u}}+\varepsilon_{t-q}
$$

$e_{t}^{u} \sim i . i . d . \Rightarrow$ no effect on dynamics of $k$

- econometrician who estimates VAR $\left\{a_{t}, k_{t}\right\}$ and ignores foresight attributes all dynamics of anticipated shock to unanticipated shock


## Blanchard-Perotti

- Legislative lags used to achieve identification
- B-P admit identification is tenuous if foresight taken seriously
- In our simple model, the VAR representation yields

$$
\begin{gathered}
k_{t}=\alpha k_{t-1}+\eta_{t}^{k}, \\
\hat{\tau}_{t}=-\kappa \delta^{2} k_{t-1}+\kappa \alpha \delta^{2} k_{t-2}+\eta_{t}^{\tau} .
\end{gathered}
$$

where $\eta_{t}^{k}=\delta^{-1} \varepsilon_{A, t}^{*}$ and $\eta_{t}^{\tau}=\delta \varepsilon_{\tau, t}^{*}$

- Use $\eta_{t+1}^{\tau}$ as instrument for agent's news at $t$
- But $\eta_{t+1}^{\tau}=\delta \varepsilon_{\tau, t+1}^{*}=\delta\left[\delta \varepsilon_{\tau, t}+\kappa \varepsilon_{A, t}\right]$
- Instrument is a mongrel shock, confounding tax news and technology


## Mountrord-Uhlig

- Ambitious-identify several shocks: taxes, spending, monetary policy, business cycle
- Use sign restrictions to address fiscal foresight
- Impose zero restrictions on response of fiscal variables over period of foresight: tax revenues cannot move for $q$ periods
- Delivers eccentric result that output falls while tax revenues do not change $\Rightarrow$ implicitly injects a sequence of unanticipated tax-rate shocks
- Sign-restrictions also likely to incorrectly other shocks, on whom the tax identification is conditional


## Ex-Ante Approach

- Reject VARs ex-ante as unable to align information sets
- Foresight implies agents' news a function of current \& future data
- In the simple analytical example

$$
\begin{aligned}
\kappa \varepsilon_{\tau, t}=k_{t} & -\left(\alpha^{-1}+\theta\right) k_{t+1}+\theta\left(\alpha^{-1}+\theta\right) k_{t+2} \\
& -\theta^{2}\left(\alpha^{-1}+\theta\right) k_{t+3}-\cdots
\end{aligned}
$$

- Ex-ante approach uses changes in revenue forecasts due to legislation to instrument for $\left\{k_{t+j}\right\}$
- Richer model: $\varepsilon_{\tau, t}$ a linear combo of all responses of endogenous variables
- Romer-Romer use narrative to classify forecasted revenue changes as "endogenous" or "exogenous"
- Need to interpret narrative method


## Formalizing the Narrative Method

- To reflect multiplicity of motivations for tax policy in Romers' narrative

$$
\begin{aligned}
\hat{\tau}_{t}= & \rho(L) \hat{\tau}_{t-1}+\underbrace{\sum_{j=-P}^{P} \mu_{j}^{C} E_{t} y_{t+j}^{C}+\sum_{j=-M}^{M} \beta_{j} E_{t} g_{t+j}}_{" R R \text { endogenous" }} \\
& +\underbrace{\sum_{j=-P}^{P} \mu_{j}^{T} E_{t} y_{t+j}^{T}+\sum_{j=-N}^{N} \gamma_{j} E_{t} s_{t+j-1}^{B}+\varepsilon_{\tau, t-q}+e_{\tau, t}^{u}}_{" R R \text { exogenous" }}
\end{aligned}
$$

## Formalizing the Narrative Method

- Specialize tax rule \& information flows to

$$
\tau_{t}=\rho \tau_{t-1}+\mu^{C} y_{t}+\xi_{t-q}+e_{\tau, t}^{u}
$$

where foresight is given by $\xi_{t-q}$ and

$$
\xi_{t-q}=\mu^{T} y_{t-q-1}+\gamma s_{t-q-1}^{B}+\varepsilon_{\tau, t-q}
$$

- Embed various specifications of tax behavior in DSGE model with capital and labor tax rates
- Simulate data \& forecasted revenues
- Estimate VARs with forecasted revenues on right-side

$$
X_{t}=C X_{t-1}+\sum_{i=0}^{24} D_{i} T_{t-i}^{u}+\sum_{i=0}^{24} F_{i} T_{t-i}^{a}+\sum_{i=1}^{6} G_{i} T_{t+i}^{a}+u_{t}
$$

## Formalizing the Narrative Method

- Alternative parametric interpretations of narrative method
(a) taxes exogenous; transfers adjust

$$
\left(\mu^{C}=0, \mu^{T}=0, \gamma_{T}=-.1, \sigma_{K}=.025, \sigma_{L}=.02\right)
$$

(b) automatic stabilizers; taxes adjust

$$
\left(\mu^{C}=1, \mu^{T}=0, \gamma_{\tau}=.05, \sigma_{K}=.025, \sigma_{L}=.02\right)
$$

(c) automatic stabilizers; response to trend; taxes adjust

$$
\left(\mu^{C}=1, \mu^{T}=.5, \gamma_{\tau}=.05, \sigma_{K}=.025, \sigma_{L}=.02\right)
$$

(d) (c) with higher relative variability of anticipated taxes

$$
\begin{aligned}
& \left(\mu^{C}=1, \mu^{T}=5, \gamma_{\tau}=.05, \sigma_{K a}=.0375, \sigma_{K u}=.0125, \sigma_{L a}=\right. \\
& \left..03, \sigma_{L u}=.01\right)
\end{aligned}
$$

- Data and forecasts come from single coherent model
- If ex-ante efficacious, should nail true effects


## Cons. Responses to Labor Taxes



## Cons. Responses to Labor Taxes






## Summary of Ex-Ante Approach

- Ex-ante approach may perform well or poorly: Conditional on how narrative approach formalized
- Narrative method of identification is not uniquely reproducible
- Different reasonable formalizations produce different conclusions
- Ex-ante approach does not model information flows: The more exogenous the forecasted revenues, the better the performance
- Connection between policy behavior and agents' information left implicit
- Difficult to integrate identification scheme into efforts to estimate DSGE models


## Appendix: Root Flipping Via Kalman

## Filter

- A very cool result: instead of using Blaschke factors, as in the paper, could obtain econometrician's information set-the VAR-using Kalman filter
- This may be surprising
- Usually we use the Kalman filter to get best linear prediction in models with latent variables
- But with non-invertibility induced by foresight, Kalman filter will not align agents and econometrician's info sets
- Kalman filter will, however, correctly recover the econometrician's info set


## Root Flipping via Kalman Filter

- Deriving the fundamental (or invertible) representation is referred to as "root flipping"
- Two ways to flip roots: Blaschke factors \& Kalman filter
- Consider the following state space representation

$$
\begin{gather*}
x_{t+1}=A x_{t}+G w_{1 t+1}  \tag{4}\\
y_{t}=C x_{t}+w_{2 t}
\end{gather*}
$$

where $\left[w_{1, t+1}^{\prime}, w_{2 t}^{\prime}\right]$ is a white noise vector with covariance matrix

$$
E\left[\begin{array}{c}
w_{1 t+1}  \tag{5}\\
w_{2 t}
\end{array}\right]\left[\begin{array}{c}
w_{1 t+1} \\
w_{2 t}
\end{array}\right]^{\prime}=\left[\begin{array}{ll}
V_{1 t} & V_{3 t} \\
V_{3 t}^{\prime} & V_{2 t}
\end{array}\right]
$$

## Root Flipping via Kalman Filter

- Two useful representations can be derived from the Kalman filter
- The first is an "innovations representation"

$$
\begin{align*}
\hat{x}_{t+1} & =A \hat{x}_{t}+K a_{t}  \tag{6}\\
y_{t} & =C \hat{x}_{t}+a_{t}
\end{align*}
$$

where $K$ is the Kalman gain, $\hat{x}_{s}$ is the optimal projection of $x_{s}$ conditional on observing $y_{s}, y_{s-1}, \ldots$, and $a_{t}$ is the innovation in predicting $y_{t}$ linearly from observing current and past $y$ 's

- The covariance matrix of the innovations is given by

$$
E a_{t} a_{t}^{\prime}=C \Sigma_{t} C^{\prime}+V_{2 t}
$$

where $\Sigma_{t}$ solves the matrix Ricatti equation

$$
\Sigma_{t+1}=A \Sigma_{t} A^{\prime}+G V_{1 t} G^{\prime}-\left(A \Sigma_{t} C^{\prime}+G V_{3 t}\right)\left(C \Sigma_{t} C^{\prime}+V_{2 t}\right)^{-1}\left(A \Sigma_{t} C+G V_{3 t}\right)^{\prime}
$$

## Root Flipping via Kalman Filter

- Can now see the invertibility condition clearly; using the lag operator $L$ and solving innovations rep for $y_{t}$ gives

$$
y_{t}=\left[I+C\left(L^{-1} I-A\right)^{-1} K\right] a_{t}
$$

- In order for $y_{t}$ (the observables) to span the same linear space as the innovations, the zeroes of the determinant:

$$
\operatorname{det}\left[I+C(z I-A)^{-1} K\right]=\frac{\operatorname{det}[z I-(A-K C)]}{\operatorname{det}(z I-A)}=0, \Rightarrow|z|<1
$$

cannot be outside the unit circle

- $z=L^{-1}$ here; this condition is equivalent to the condition in LWY
- the zeros of $\operatorname{det}[z I-(A-K C)]$ are the eigenvalues of $A-K C$


## Root Flipping via Kalman Filter

- The second representation that is useful to consider is the whitening filter

$$
\begin{gather*}
\hat{x}_{t+1}=(A-K C) \hat{x}_{t}+K y_{t}  \tag{7}\\
a_{t}=y_{t}-C \hat{x}_{t}
\end{gather*}
$$

- a whitening filter takes a sequence of $y$ 's and gives as output a sequence of $a$ 's that are serially uncorrelated
- can see why the invertibility condition is crucial
- if the eigenvalues of $A-K C$ are not all inside the unit circle, then (7) is not a stationary process


## Root Flipping via Kalman Filter

- Consider the following ARMA process from LWY

$$
k_{t}=\alpha k_{t-1}-\kappa\left\{\varepsilon_{\tau, t-1}+\theta \varepsilon_{\tau, t}\right\}
$$

- A state space formulation for this process is given by

$$
\begin{gather*}
x_{t}=-(\kappa \theta)^{-1} k_{t}-\varepsilon_{\tau, t}, y_{t}=-(\kappa \theta)^{-1} k_{t} \\
x_{t+1}=\alpha x_{t}+\left[\alpha+\theta^{-1}\right] \varepsilon_{\tau, t} \\
y_{t}=x_{t}+\varepsilon_{\tau, t} \tag{8}
\end{gather*}
$$

and $V_{1}=V_{2}=V_{3}$. We assume that the initial state is known ( $\Sigma_{t_{0}}=0$ ), which implies that the Kalman gain is given by $K_{t}=G$ for all $t$ and then

$$
A-K C=\alpha-\alpha-\theta^{-1}=\left|\theta^{-1}\right|>1
$$

The root of $z-A+K C$ is outside the unit circle and the innovations of ( 8 ) ( $\varepsilon_{\tau, t}$ ) do not span the same space as the observables and therefore (8) cannot be the innovations representation

## Root Flipping via Kalman Filter

- Time invariance of the Kalman filter requires two primary assumptions:

1. The pair $\left(A^{\prime}, C^{\prime}\right)$ is stabilizable. A pair $(A, C)$ is stabilizable if $y^{\prime} C=0$ and $y^{\prime} A=\lambda y^{\prime}$ for some complex number $\lambda$ and some complex vector $y$ implies that $|\lambda|<1$ or $y=0$
2. The pair $(A, G)$ is detectable. The pair $(A, G)$ is detectable if $G^{\prime} y=0$ and $A y=\lambda y$ for some complex number $\lambda$ and some complex vector $y$ implies that $|\lambda|<1$ or $y=0$

- Given (1) \& (2), iterations on the matrix Ricatti equation converge as $t \rightarrow \infty$, starting from any semi-positive definite matrix $\Sigma_{t_{0}}$
- This then implies a time invariant Kalman gain $K$
- And $A-K C$ is a stable matrix with eigenvalues less than unity in modulus


## Root Flipping via Kalman Filter

- Returning to the example, both assumptions hold for the non-invertible ARMA process
- Both conditions yield $\alpha=\lambda<1$ or $y=0$-implying there exists a $K$ such that $A-K C$ is less than one
- Note that the Ricatti equation and time-invariant solution are given by

$$
\begin{gathered}
\sigma_{t+1}=\alpha^{2} \sigma_{t}+\left(\alpha+\theta^{-1}\right)^{2}+\frac{\left(\alpha \sigma_{t}+\alpha+\theta^{-1}\right)^{2}}{1+\sigma_{t}} \\
\sigma_{\infty}=\frac{1-\theta^{2}}{\theta^{2}}
\end{gathered}
$$

This gives the Kalman gain as
$K=\left(A \Sigma_{\infty} C^{\prime}+G V_{3 t}\right)\left(C \Sigma_{\infty}+V_{2 t}\right)^{-1}=\frac{\alpha \sigma_{\infty}+\alpha+\theta^{-1}}{1+\sigma_{\infty}}=\alpha+\theta$
and now $A-K C=-\theta<1$

## Root Flipping via Kalman Filter

- The representation gives the innovation as

$$
\begin{aligned}
-(\kappa \theta)^{-1} k_{t} & =\left[1+\frac{(\alpha+\theta)}{\left(L^{-1}-\alpha\right)}\right] a_{t} \\
& =\left[\frac{L^{-1}+\theta}{L^{-1}-\alpha}\right] a_{t} \\
& =\left[\frac{1+\theta L}{1-\alpha L}\right] a_{t} \\
(1-\alpha L) k_{t}=-\kappa(1+\theta L) \theta a_{t} &
\end{aligned}
$$

- This is equivalent to equation (13) of LWY
- It implies that $\theta a_{t}=\varepsilon_{t}^{*}$, so the impulse response function using the Kalman filter must be normalized by the standard deviation of $a_{t}$

