Optimal Policy Without Rational Expectations: A Sufficient Statistic Solution

Jonathan Adams – University of Florida University of Virginia – January 15 2025

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- Lack of generality is a problem: no consensus on how expectations are formed (beyond FIRE fails), precisely how they affect the real economy, etc.

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- The policymaker does *not* need to know:
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 - The entire economic model
- The policymaker only needs to:
 - Measure peoples' expectations
 - Know how decisions (equilibrium conditions) are directly distorted by non-rational expectations

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 - Cannot recover FIRE exactly
 - But the belief distortion is still a sufficient statistic for the optimal policy!
- Work through simple examples for both cases

General Framework

• General model:

$$B_{X1}\mathbb{E}_{t}^{b}\left[X_{t+1}\right] = B_{X0}X_{t} + B_{Y}Y_{t} + B_{G}G_{t} \tag{1}$$

- $\mathbb{E}_t^b[\cdot]$: behavioral expectation of type b
- X_t : endogenous variables
- Y_t: exogenous variables
- G_t : policy variables

• A behavioral expectations equilibrium:

- 1. X_t , Y_t , and G_t satisfy the equilibrium condition (1)
- 2. Y_t , X_t and G_t are stationary, linear in the history of shocks $\{\omega_{t-j}\}_{j=0}^{\infty}$
- 3. G_t satisfies a policy rule
- ullet For now: assume FIRE equilibrium X_t^* is welfare-maximizing, unique

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- See Adams (2023) for technical details

Belief Distortions

- $\mathbb{E}_t[\cdot]$ (with no *b* specified) denotes the *rational expectation*
- Define the **belief distortion** as

$$\mathbb{D}_t^b[X_{t+1}] \equiv \mathbb{E}_t^b[X_{t+1}] - \mathbb{E}_t[X_{t+1}]$$

- ullet In a model, it is specific to the type b of behavioral expectations
- In the data, requires measuring agents' expectations $\mathbb{E}_t^b[X_{t+1}]$, and estimating the rational expectation $\mathbb{E}_t[X_{t+1}]$

What Must Optimal Policy Do?

Lemma

If there is a time series of policy instruments G_t such that the non-rational equilibrium is consistent with the policy-less FIRE equilibrium, then G_t satisfies

$$B_{X1}\mathbb{D}_t^b[X_{t+1}]=B_GG_t$$

Proof Outline:

• In the FIRE equilibrium with $G_t = 0$, endogenous vector X_t^* satisfies:

$$B_{X1}\mathbb{E}_t\left[X_{t+1}^*\right] = B_{X0}X_t^* + B_YY_t$$

Subtract from the non-rational model to get:

$$B_{X1}\mathbb{E}_{t}^{b}\left[X_{t+1}\right] - B_{X1}\mathbb{E}_{t}\left[X_{t+1}^{*}\right] = B_{X0}(X_{t} - X_{t}^{*}) + B_{G}G_{t}$$

• Impose $X_t = X_t^*$, and rearrange.

Sentiment Spanning: Definition

- What policy instruments are enough to recover FIRE?
- Some notation:
 - B_{C1} is submatrix of B_{X1} corresponding to control variables (there is no belief distortion about pre-determined state variables)
 - $P_G \equiv B_G (B_G' B_G)^{-1} B_G'$ is projection onto column space of B_G .

Condition (Sentiment Spanning)

The macroeconomic model defined in (1) is said to satisfy Sentiment Spanning if

$$(I-P_G)\,B_{C1}=0$$

Sentiment Spanning in Practice

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- Policymaker does not need to know the whole model to evaluate SS! Needs to know:
 - How expectations affect decisions (B_{C1})
 - How policy instruments distort economy (B_G)

Optimal Policy: The Sufficient Statistic

Theorem

If a model satisfies Sentiment Spanning, then the policy rule

$$G_t^{\dagger} = (B_G' B_G)^{-1} B_G' B_{C1} \mathbb{D}_t^b [X_{t+1}^C]$$
 (2)

recovers the FIRE equilibrium.

- The belief distortion $\mathbb{D}_t^b[X_{t+1}^C]$ is a sufficient statistic!
- ullet Why does Sentiment Spanning matter? Invert the Lemma $B_{X1}\mathbb{D}_t^b[X_{t+1}]=B_GG_t$

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- Optimal policy: tax capital when agents are overly optimistic about future returns

Example 1: Decentralized Equilibrium Conditions

• Policymakers have light information requirements:

Euler Equation:
$$\tau_t = \sigma c_t + \mathbb{E}_t^b [-\sigma c_{t+1} + \overline{R} r_{t+1}]$$
 Labor Supply:
$$w_t = \sigma c_t + \eta n_t$$

Production Function:
$$y_t = a_t + \alpha k_{t-1} + (1 - \alpha)n_t$$

Capital Demand:
$$r_t = y_t - k_{t-1}$$

Labor Demand:
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Resource Constraint:
$$\overline{Y}y_t = \overline{C}c_t + \overline{K}(k_t - (1-\delta)k_{t-1})$$

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• Optimal policy: $\tau_t^{\dagger} = \mathbb{D}_t^b[-\sigma c_{t+1} + \overline{R}r_{t+1}]$

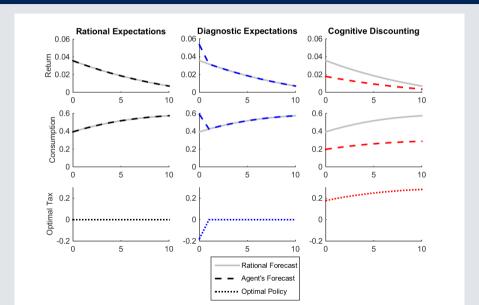
Example 1: Types of Behavioral Expectations

Rational Expectations:
$$\mathbb{E}_t^{RE}[x_{t+1}] = \mathbb{E}_t[x_{t+1}]$$

Diagnostic Expectations:
$$\mathbb{E}_t^{DE}[x_{t+1}] = (1 + \theta^{DE})\mathbb{E}_t[x_{t+1}] - \theta^{DE}\mathbb{E}_{t-1}[x_{t+1}]$$

Cognitive Discounting:
$$\mathbb{E}_t^{CD}[x_{t+1}] = \theta^{CD}\mathbb{E}_t[x_{t+1}]$$

Example 1: Response of Expectations to a Productivity Shock



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 - ... but with only monetary, FIRE cannot be recovered
- Both cases: raise interest rates when agents misperceive the economy to be running hot

New Keynesian Phillips Curve:
$$\psi f_t = \kappa y_t - \pi_t - z_t^{PC} + \beta \mathbb{E}_t^b[\pi_{t+1}]$$
 Euler Equation:
$$i_t = -\sigma y_t - z_t^{EE} + \mathbb{E}_t^b[\sigma y_{t+1} + \pi_{t+1}]$$

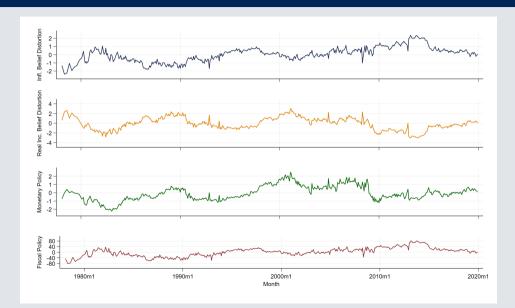
Expectation components of optimal policy are:

$$\hat{f}_t^{\dagger} = \frac{\beta}{\psi} \mathbb{D}_t^b \left[\hat{\pi}_{t+1} \right] \qquad \qquad \hat{i}_t^{\dagger} = \mathbb{D}_t^b \left[\sigma y_{t+1} + \pi_{t+1} \right]$$

Implementation:

- Measure agents' expectations $\mathbb{E}_t^b[\cdot]$,
- Estimate the rational expectation, e.g. with a VAR (Adams and Barrett 2024)

Example 2: Estimated Belief Distortions and Implied Policies



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 - 3. What if expectation formation is endogenous?
- ... intuition goes through, although implementation may change

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- ... but they still do not need to know how expectations are formed!

Welfare Objective

- First-best equilibrium: X_t^* , with FIRE-optimal policy G_t^*
- Policymakers with no information commit to a policy rule (Rottemburg and Woodford 1997)
- Minimize quadratic loss for some *W*:

$$\min \mathbb{E}\left[(X_t - X_t^*)'W(X_t - X_t^*)\right]$$

Theorem

The constrained-optimal policy rule is

$$G_{t}^{\dagger} = \underbrace{B_{G}^{+} P_{W} B_{C1} \mathbb{D}_{t}^{b} [X_{t+1}^{C}]}_{expectation\ component} + G_{t}^{RE}$$

→ matrix details

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- Expectation component $B_G^+ P_W B_{C1} \mathbb{D}_t^b [X_{t+1}^C]$ has the same information requirements as the original case with sentiment spanning
- if you already have the FIRE optimal policy, adding the response to non-rational expectations requires no additional modeling assumptions, only measuring the belief distortion!

Example 3: Optimal Monetary Policy Alone in BNK

New Keynesian Phillips Curve: $0 = \kappa y_t - \pi_t - z_t^{PC} + \beta \mathbb{E}_t^k [\pi_{t+1}]$ Euler Equation: $i_t = -\sigma y_t - z_t^{EE} + \mathbb{E}_t^b [\sigma y_{t+1} + \pi_{t+1}]$

Expectation component of optimal policy is:

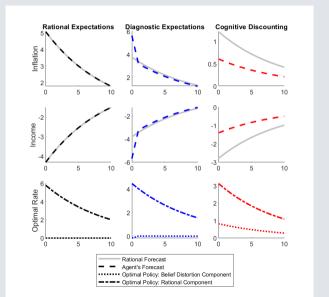
$$\hat{i}_t^{\dagger} - i_t^{RE} = \sigma \mathbb{D}_t^b[y_{t+1}] + \left(1 - \beta \frac{b_{\pi} \kappa \sigma}{b_{\pi} \kappa^2 + b_y}\right) \mathbb{D}_t^b[\pi_{t+1}]$$

which cannot recover FIRE without an additional tool.

If $\left(1 - \beta \frac{b_{\pi}\kappa\sigma}{b_{\pi}\kappa^2 + b_y}\right) > 0$, raise rates when agents misperceive economy is "running hot".

▶ Speedy Conclusion

Example 3: Response of Expectations to a Cost-Push Shock



What if Belief Distortions are Measured with Error?

ullet Policymaker's observation D_t of the belief distortion is

$$D_t = \xi \mathbb{D}_t^b[X_{t+1}] + v_t$$

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• Form the *policymaker's* nowcast of the belief distortion $\mathbb{D}_t^b[X_{t+1}^C]$ conditional on info. set Ω_t (D_t and other observables):

$$\hat{D}_t = \mathbb{E}[\mathbb{D}_t^b[X_{t+1}^C]|\Omega_t]$$

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• Theorem The constrained-optimal policy rule is

$$G_t^{\dagger} = B_G^+ P_W B_{C1} \hat{D}_t + G_t^{RE}$$

... same as the solution without Sentiment Spanning, except using \hat{D}_t !

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- Now let the operator $\mathbb{E}^b_t[\cdot;\mathcal{G}]$ depend on the policy rule \mathcal{G}
- Return to simple case: sentiment spanning holds, FIRE is optimal

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- ... but Theorem 1 statement is false, because of $\mathbb{D}_t^b[X_{t+1}^c;\mathcal{G}]$ nonlinearity
 - G_t^{\dagger} may not be unique
 - G_t^{\dagger} may not even exist! (example in sec. 6.2.2)

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Theorem

The constrained-optimal policy rule is

$$G_t^{\dagger} = B_G^+ P_W B_{C1} \mathbb{D}_t^b [X_{t+1}^C] + B_G^+ P_W B_{X1} \mathbb{E}_t [X_{t+1} - X_{t+1}^*] + G_t^*$$

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$$B_G^+ \equiv (B_G' B_G)^{-1} B_G'$$
, $P_W \equiv B_G \left(B_G' \tilde{W} B_G \right)^{-1} B_G' \tilde{W}$, $\tilde{W} \equiv \left(B_{X0}^{-1} \right)' W B_{X0}^{-1}$

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 - 2. Economic component: optimal policy for FIRE model

▶ back

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 - Optimal rule unchanged; lose existence/uniqueness from the main theorem

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