# **BAYESIAN ESTIMATION**

CALIBRATION EXAMPLE

Tools for Macroeconomists: The essentials

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Calibration example

#### A STATE-OF-THE-ART LABOR MARKET MODEL



#### SIMPLIFIED SEARCH AND MATCHING MODEL

Model for understanding worker flows on the labor market Main idea: labor market characterized by frictions

- finding (the right) job takes time and resources
- $\cdot \rightarrow$  co-existence of unemployed (*U*) and vacant jobs (*V*)
- $\cdot \rightarrow$  not all workers (firms) find (fill) jobs
  - probability of finding a job: f = M/U
  - probability of filling a job: q = M/V
  - M are total hires:  $M = mU^{\mu}V^{1-\mu}$ 
    - matching function summarizes labor market frictions

#### TWO MODEL EQUATIONS

Unemployment dynamics
 (unemployment tomorrow = unlucky unemployed and fired workers)

$$U_{t+1} = (1 - f_t)U_t + s(1 - U_t)$$

- · labor force normalized to 1
- s is separation rate (constant for now)
- Optimal recruitment by firms (costs = expected benefits benefits)

$$\kappa = q_t \beta \mathbb{E} J_{t+1}$$

•  $J_t$  is value of a job for the firm

## VALUE OF A JOB FOR THE FIRM

What's in it for the firm to hire a worker?

$$J_t = y_t - w_t + \beta(1 - s)\mathbb{E}J_{t+1}$$

- y output per worker (exogenous process)
- w wage paid to worker
  - assume wages are a constant share  $(\alpha)$  of output

#### PUTTING EVERYTHING TOGETHER

The model boils down to 2 equations for 2 model variables

$$U_{t+1} = (1 - mU_t^{\mu-1}V_t^{1-\mu})U_t + s(1 - U_t)$$

$$\frac{\kappa}{mU_t^{\mu}V_t^{-\mu}} = \mathbb{E}\beta \left( y_{t+1}(1 - \alpha) + (1 - s)\frac{\kappa}{mU_{t+1}^{\mu}V_{t+1}^{-\mu}} \right)$$

We need to pick values for the 6 parameters:

• m, 
$$\mu$$
, s,  $\kappa$ ,  $\beta$ ,  $\alpha$ 

### **CALIBRATING OUR MODEL**

# "Easy" choices:

- $\beta$ : time preference parameter (frequency of model)
- s: separation rate (observed in the data)

#### "Harder" choices:

- $\cdot$   $\mu$ : elasticity of matching function (estimated from data)
- m: matching function level (normalization)

#### "Hard" choices:

- $\alpha$ : could interpret it as labor share in income (data?)
- $\kappa$ : could interpret it as (in-)direct hiring costs (data?)

### POSSIBLE ISSUES WITH ABOVE CALIBRATION

#### Obvious concerns:

- $\cdot \alpha$  not really labor share (model has no capital)
- $\cdot$   $\kappa$  literally a per-period resource cost of an open vacancy

#### Less obvious issues:

- even if we're happy with all our targets
- the calibration may be inconsistent (FOC's may not hold)
- or implied steady state values don't make sense
- way out?
  - use steady state values of variables as targets
  - back out implied parameter values

# Calibration example

TAKING STOCK

#### TAKING STOCK

# Calibration example

- instead of directly setting parameter values
- · it is often useful to set steady states of certain variables
- and then back out the implied parameter value that achieves such a steady state
- note that this may not always be possible for some combinations of parameters

