# Neural Network Application to Stochastic Overlapping Generations Economies

### Kevin Mott

January 28, 2022

# **Economic Model**

Notation:  $z_i^t$  indicates value of economic variable z at time t for the  $i^{\text{th}}$  youngest agent.

Economic Variable:	Notation:
Endowment income	$\omega_i^t$
Consumption	$c_i^t$
Cash-on-hand	$x_i^t$
Equity holdings	$e_i^t$
Equity price	$p^t$
Equity dividend	$\delta^t$
Total resources	$w^t = \delta^t + \sum_{i=1}^L \omega_i^t$
Shock	$s^t$

Intertemporal budget constraints:

$$c_1^t = \omega_1^t - p^t e_1^t - q^t b_1^t$$

$$c_i^t = \omega_i^t + (p^{t+1} + \delta^{t+1}) e_{i-1}^{t-1} + b_{i-1}^{t-1} - p^t e_i^t - q^t b_i^t \quad \text{for } i \in \{2, \dots, L-1\}$$

$$c_L^t = \omega_L^t + (p^{t+1} + \delta^{t+1}) e_{L-1}^{t-1} + b_{L-1}^{t-1}$$

Leading to the consumer's problem:

$$\max_{\left(e_{i}^{t+\tau}\right)_{\tau=0}^{L-1},\left(b_{i}^{t+\tau}\right)_{\tau=0}^{L-1}} \left\{ u\left(c_{i}^{t}\right) + \sum_{\tau=1}^{L-1} \beta^{\tau} \mathbb{E}_{t}\left[u\left(c_{i}^{t+\tau}\right)\right] \right\} \quad \text{s.t. Budget Constraints}$$

Market clearing conditions:

$$0 = \sum_{i=1}^{L-1} b_i^t$$
 
$$1 = \sum_{i=1}^{L-1} e_i^t$$

Theory tells us the relevant state variable at time t is

$$\sigma^t := \left( (e_i^t)_{i=1}^{L-1}, (b_i^t)_{i=1}^{L-1}, s^t \right)$$

# **Neural Network Computation**

Forward time iteration to create Ergodic grid. Starting at the solution to the equity-only deterministic steady state:  $\sigma^0 = (\bar{e}, \bar{b}, s^0)$ .

#### Network Architecture

At time t:

• Input:  $\sigma^t = ((e_i^t)_{i=1}^{L-2}, (b_i^t)_{i=2}^{L-1}, s^t) \in \mathbb{R}^{2L-3}$ 

• Output:  $y^t(\sigma^t) := ((e_i^t)_{i=1}^{L-2}, (b_i^t)_{i=1}^{L-2}, p^t, q^t) \in \mathbb{R}^{2L-2}$ 

So we see that each period, the neural network takes as input the output of last period. Run this forward for T periods.

Use market clearing for  $e_{L-1}^t$  and  $b_{L-1}^t$  in each period t.

### Loss Function

I use Euler equation errors as loss function. At time t:

$$\mathcal{E}^{t} := \sum_{i=1}^{L-1} \left| \frac{(u')^{-1} \left[ \frac{\beta \mathbb{E}_{t} \left[ u'(c_{i+1}^{t+1}) \right]}{q^{t}} \right]}{c_{i}^{t}} - 1 \right| + \sum_{i=1}^{L-1} \left| \frac{(u')^{-1} \left[ \frac{\beta \mathbb{E}_{t} \left[ (p^{t+1} + \delta^{t+1}) u'(c_{i+1}^{t+1}) \right]}{p^{t}} \right]}{c_{i}^{t}} - 1 \right|$$

## Training

If the parameters of the neural network are  $\Gamma$ , the optimization scheme is:

$$\min_{\Gamma} \sum_{t=0}^{T} \mathcal{E}^{t}$$