# Derivations of a working example based on Brunnermeier and Sannikov (2014)

Adrien d'Avernas[*], Valentin Schubert[†] and Quentin Vandeweyer[‡]

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   This document presents a first working example for the Matlab toolbox from d'Avernas et al. (2020)[1]. We derive the model and relate the different expressions to the equations needed to solve the model numerically using the toolbox.

In this first example, we present a general extension of Brunnermeier and Sannikov (2014) where two agents have Epstein and Zin (1989) utility functions and aggregate volatility is time-varying. The framework can easily be modified to any other general equilibrium framework with $n$-agents and two state variables.

**Preferences**   There are two types of agents: households $h \in H$ and intermediaries $i \in I$. Both agents have stochastic differential utility, as developed by Duffie and Epstein (1992). The utility of agent $j$ over his consumption process $c_t^j$ is defined as

$$U_t^j = \mathbb{E}_t \left( \int_t^\infty f\left(c_s^j, U_s^j\right) ds \right).$$

The function $f_j(c, u)$ is a normalized aggregator of consumption and continuation value in each period defined as

$$f(c, U) = \frac{1 - \gamma}{1 - 1/\zeta} U \left[ \left( \frac{c}{((1 - \gamma)U)^{1/(1-\gamma)}} \right)^{1 - 1/\zeta} - \rho \right]$$

where $\rho$ is the rate of time preference, $\gamma$ is the coefficient of relative risk aversion, and $\zeta$ determines the elasticity of intertemporal substitution. Each agent chooses its optimal consumption $c_t^j$, investment $\iota_t^j$, and portfolio weight $w_t^j$ on capital holdings

[*]Stockholm School of Economics
[†]Stockholm School of Economics
[‡]European Central Bank and Sciences Po
[1]Available for download under https://github.com/adavernas/toolbox

in order to maximize discounted infinite life time expected utilities $U_t^j$. At any time, the following budget constraint has to be satisfied:

$$\frac{dn_t^j}{n_t^j} = \left(\left(1 - w_t^j\right)r_t + w_t^j\mu_t^{r,j} - \mathfrak{c}_t^j\right)dt + w_t^j\sigma_t^{q,\sigma}dZ_t^\sigma + w_t^j\left(\sigma_t + \sigma_t^{q,k}\right)dZ_t^k,$$

where $n_t^j$ is the wealth of agent $j$, $\mathfrak{c}_t^j = c_t^j/n_t^j$ his consumption rate, and the portfolio weight $w_t^j$ are choice variables. $Z_t^\sigma$ and $Z_t^k$ are two standard Brownian motions that hit aggregate volatility and capital growth, respectively.

**Technology**    The production technology in the economy is given by:

$$y_t^j = \left(a^j - \iota_t^j\right)k_t^j$$

and

$$\frac{dk_t}{k_t} = \mu_t^k dt + \sigma_t dZ_t^k,$$

where $\mu_t^k$ is given by

$$\mu_t^k = \psi_t\Phi(\iota_t^i) + (1 - \psi_t)\Phi(\iota_t^h),$$

$\Phi(\cdot)$ is a concave investment function, and $\psi_t$ is the share of capital in the hands of intermediaries:

$$\psi_t \equiv \frac{w_t^i n_t^i}{w_t^i n_t^i + w_t^h n_t^h}.$$

In this model, we work with the following functional form for the investment function:

$$\Phi(\iota_t^j) = log(1 + \kappa_p\iota_t^j)/\kappa_p - \delta^j$$

The price of a unit of capital is $q_t$. The volatility of capital returns follows a diffusion:

$$\frac{d\sigma_t}{\sigma_t} = \kappa\left(\sigma_t - \bar{\sigma}\right)dt + \varsigma dZ_t^\sigma.$$

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The stochastic law of motion of $q_t$ follows:

$$\frac{dq_t}{q_t} = \mu_t^q dt + \sigma_t^{q,\sigma} dZ_t^\sigma + \sigma_t^{q,k} dZ_t^k.$$

The variables $\mu_t^q$, $\sigma_t^{q,k}$, and $\sigma_t^{q,\sigma}$ are to be determined endogenously. We can use Ito's lemma to write the process of the value of capital:

$$\frac{d(q_t k_t^j)}{q_t k_t^j} = \left(\mu_t^k + \mu_t^q + \sigma_t \sigma_t^{q,k}\right) dt + \sigma_t^{q,\sigma} dZ_t^\sigma + \left(\sigma_t + \sigma_t^{q,k}\right) dZ^k.$$

Hence, the return on the physical asset is:

$$dr_t^j = \underbrace{\left(\frac{a^j - \iota_t^j}{q_t} + \mu_t^k + \mu_t^q + \sigma_t \sigma_t^{q,k}\right)}_{\mu_t^{r,j}} dt + \sigma_t^{q,\sigma} dZ_t^\sigma + \left(\sigma_t + \sigma_t^{q,k}\right) dZ^k.$$

**Solving the HJB**   We will guess and verify that the homotheticity of preferences allows us to write the value function for agents of type $j$ as:

$$U\left(n_t^j, \xi_t^j\right) = \frac{\left(n_t^j\right)^{1-\gamma} \xi_t^j}{1-\gamma},$$

where variable $\xi_t^j$ follows

$$\frac{d\xi_t^j}{\xi_t^j} = \mu_t^{\xi,j} dt + \sigma_t^{\xi,\sigma,j} dZ_t^\sigma + \sigma_t^{\xi,k,j} dZ_t^k.$$

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We can write the HJB equation corresponding to the problem of agent $j$ as

$$
\begin{aligned}
0 = \max_{\mathfrak{c}_t^j, \iota_t^j, w_t^j} & f\left(\mathfrak{c}_t^j n_t^j, U_t^j\right) \\
& + \left(\left(1 - w_t^j\right)r_t + w_t^j \mu_t^{r,j} - \mathfrak{c}_t^j\right)n_t^j U_n(n_t^j, \xi_t^j) + \mu_t^{\xi,j}\xi_t^j U_\xi(n_t^j, \xi_t^j) \\
& + \frac{1}{2}\left[\left(w_t^j \sigma_t^{q,\sigma} n_t^j\right)^2 + \left(w_t^j\left(\sigma_t + \sigma_t^{q,k}\right)n_t^j\right)^2\right] U_{nn}(n_t^j, \xi_t^j) \\
& + \frac{1}{2}\left[\left(\sigma^{\xi,\sigma,j}\xi_t^j\right)^2 + \left(\sigma^{\xi,k,j}\xi_t^j\right)^2\right] U_{\xi\xi}(n_t^j, \xi_t^j) \\
& + \left[w_t^j \sigma_t^{q,\sigma} n_t^j \sigma_t^{\xi,\sigma,j}\xi_t + w_t^j\left(\sigma_t + \sigma_t^{q,k}\right)n_t^j \sigma_t^{\xi,k,j}\xi_t^j\right] U_{n\xi}(n_t^j, \xi_t^j).
\end{aligned}
$$

Substituting in the guess, the HJB becomes

$$
\begin{aligned}
0 = \max_{\mathfrak{c}_t^j, \iota_t^j, w_t^j} & \frac{1}{1 - 1/\zeta}\left[\frac{\left(\mathfrak{c}_t^j\right)^{1-1/\zeta}}{\left(\xi_t^j\right)^{\frac{1-1/\zeta}{1-\gamma}}} - \rho\right] + (1 - w_t^j)r_t + w_t^j \mu_t^{r,j} - \mathfrak{c}_t^j + \frac{\mu^{\xi,j}}{1-\gamma} \\
& - \frac{\gamma}{2}\left(w_t^j \sigma_t^{q,\sigma}\right)^2 - \frac{\gamma}{2}\left(w_t^j \sigma_t + w_t^j \sigma_t^{q,k}\right)^2 + w_t^j \sigma_t^{q,\sigma}\sigma_t^{\xi,\sigma,j} + w_t^j\left(\sigma_t + \sigma_t^{q,k}\right)\sigma_t^{\xi,k,j}.
\end{aligned}
$$

**Optimality Conditions** The first order conditions with respect to $\mathfrak{c}_t^j, \iota_t^j$, and $w_t^j$ are given by

$$
\left(\mathfrak{c}_t^j\right)^{-1/\zeta} = \left(\xi_t^j\right)^{\frac{1-1/\zeta}{1-\gamma}},
$$

$$
1/q_t = \Phi_\iota(\iota_t),
$$

$$
\mu_t^{r,j} - r_t - \gamma w_t^j\left(\sigma_t^{q,\sigma}\right)^2 - \gamma w_t^j\left(\sigma_t + \sigma_t^{q,k}\right)^2 + \sigma_t^{q,\sigma}\sigma_t^{\xi,\sigma,j} + \left(\sigma_t + \sigma_t^{q,k}\right)\sigma_t^{\xi,k,j} = 0.
$$

Plugging in the optimality conditions in the HJB gives:

$$
0 = \frac{1}{1 - 1/\zeta}\left(\mathfrak{c}_t^j - \rho\right) + r_t - \mathfrak{c}_t^j + \frac{\gamma}{2}\left(w_t^j \sigma_t^{q,\sigma}\right)^2 + \frac{\gamma}{2}\left(w_t^j \sigma_t + w_t^j \sigma_t^{q,k}\right)^2 + \frac{\mu^{\xi,j}}{1-\gamma}.
$$

**Market Clearing Conditions** We start by providing the definition of such an equilibrium in the state variables $\{\eta_t, \sigma_t\}$, where $\eta_t$ is defined as the share of wealth in the hands of the intermediaries:

$$\eta_t = \frac{n_t^i}{n_t^h + n_t^i} = \frac{n_t^i}{q_t k_t}.$$

Then, we can use the market clearing condition for consumption to find $q_t$. Market clearing for consumption dictates that consumption from both types of agents equals the surplus from the production technology:

$$\mathfrak{c}_t^i n_t^i + \mathfrak{c}_t^h n_t^h = (a^i - \iota_t^i)k_t^i + (a^h - \iota_t^h)k_t^h$$

$$\frac{\mathfrak{c}_t^i n_t^i}{q_t k_t} + \frac{\mathfrak{c}_t^h n_t^h}{q_t k_t} = \frac{(a^i - \iota_t^i)k_t^i}{q_t k_t} + \frac{(a^h - \iota_t^h)k_t^h}{q_t k_t}$$

$$\mathfrak{c}_t^i \eta_t + \mathfrak{c}_t^h(1 - \eta_t) = \frac{(a^i - \iota_t^i)n_t^i k_t^i}{n_t^i q_t k_t} + \frac{(a^h - \iota_t^h)n_t^h k_t^h}{n_t^h q_t k_t}$$

$$\mathfrak{c}_t^i \eta_t + \mathfrak{c}_t^h(1 - \eta_t) = \frac{(a^i - \iota_t^i)\eta_t k_t^i}{n_t^i} + \frac{(a^h - \iota_t^h)(1 - \eta_t)k_t^h}{n_t^h}$$

$$\mathfrak{c}_t^i \eta_t + \mathfrak{c}_t^h(1 - \eta_t) = (a^i - \iota_t^i)\eta_t w_t^i/q_t + (a^h - \iota_t^h)(1 - \eta_t)w_t^h/q_t$$

Now note that

$$\psi_t = \frac{w_t^i n_t^i}{w_t^i n_t^i + w_t^h n_t^h} = w_t^i \eta_t,$$

So finally, we have

$$\left(\mathfrak{c}_t^i \eta_t + \mathfrak{c}_t^h(1 - \eta_t)\right) q_t = \psi_t(a^i - \iota_t^i) + (1 - \psi_t)(a^h - \iota_t^h)$$

The market clearing condition for capital allows us to identify $r_t$:

$$k_t^i + k_t^h = k_t$$
$$\frac{k_t^i}{k_t} + \frac{k_t^h}{k_t} = 1$$
$$\frac{k_t^i n_t^i q_t}{n_t^i q_t k_t} + \frac{k_t^h n_t^h q_t}{n_t^h q_t k_t} = 1$$
$$w_t^i \eta_t + w_t^h (1 - \eta_t) = 1.$$

Further, using our definition of $\eta_t$ and Ito's lemma, we can derive the law of motion of $\eta_t$ as:

$$\frac{d\eta_t}{\eta_t} = \left( r_t + w_t^i(\mu_t^{r,j} - r_t) - \mathfrak{c}_t^i - \mu_t^k - \mu_t^q + (\sigma_t)^2 + (\sigma_t^{q,k})^2 + (\sigma_t^{q,\sigma})^2 \right.$$
$$\left. + \sigma_t \sigma_t^{q,k} - w_t^i(\sigma_t^{q,k})^2 - w_t^i \sigma_t^{q,k}(\sigma_t + \sigma_t^{q,k}) \right) dt$$
$$+ \left( w_t^i - 1 \right) \sigma_t^{q,\sigma} dZ_t^\sigma + \left( w_t^i - 1 \right) \left( \sigma_t + \sigma_t^{q,k} \right) dZ^k$$

By applying Ito's lemma, we can find $\sigma_t^{q,\sigma}$, $\sigma_t^{q,k}$, $\sigma_t^{\xi,\sigma,j}$, $\sigma_t^{\xi,k,j}$, and $\mu_t^q$ from:

$$q(\sigma_t, \eta_t) \sigma_t^{q,\sigma} = q_\sigma(\sigma_t, \eta_t) \varsigma \sigma_t + q_\eta(\sigma_t, \eta_t) \left( w_t^i - 1 \right) \sigma_t^{q,\sigma} \eta_t,$$
$$q(\sigma_t, \eta_t) \sigma_t^{q,k} = q_\eta(\sigma_t, \eta_t) \left( w_t^i - 1 \right) \left( \sigma_t + \sigma_t^{q,k} \right) \eta_t,$$
$$\xi^j(\sigma_t, \eta_t) \sigma_t^{\xi,\sigma,j} = \xi_\sigma^j(\sigma_t, \eta_t) \varsigma \sigma_t + \xi_\eta^j(\sigma_t, \eta_t) \left( w_t^i - 1 \right) \sigma_t^{q,\sigma} \eta_t,$$
$$\xi^j(\sigma_t, \eta_t) \sigma_t^{\xi,k,j} = \xi_\eta^j(\sigma_t, \eta_t) \left( w_t^i - 1 \right) \left( \sigma_t + \sigma_t^{q,k} \right) \eta_t,$$
$$q(\sigma_t, \eta_t) \mu_t^q = q_\sigma(\sigma_t, \eta_t) \mu_t^\sigma \sigma_t + q_\eta(\sigma_t, \eta_t) \mu_t^\eta \eta_t + \frac{1}{2} q_{\sigma\sigma}(\sigma_t, \eta_t) \left( \varsigma \sigma_t \right)^2$$
$$+ \frac{1}{2} q_{\eta\eta}(\sigma_t, \eta_t) \left[ \left( (w_t^i - 1) \sigma_t^{q,\sigma} \eta_t \right)^2 + \left( (w_t^i - 1)(\sigma_t + \sigma_t^{q,k}) \eta_t \right)^2 \right]$$
$$+ q_{\sigma\eta}(\sigma_t, \eta_t) \varsigma \sigma_t \left( w_t^i - 1 \right) \sigma_t^{q,\sigma} \eta_t.$$

**Linking the model to the code**    We start by collecting parameters and variables. The model parameters are reported in table 1. The parameters have to be specified (as well as given a value) in the section *Parameters* in the file `model.m` of the toolbox.

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**Table 1:** Model parameters

| Parameter | Definition |
|-----------|------------|
| $\gamma^j$ | relative risk aversion |
| $\zeta^j$ | intertemporal elasticity of substitution |
| $\rho$ | discount rate |
| $a^j$ | productivity |
| $\kappa_p$ | investment costs |
| $\kappa$ | drift of volatility |
| $\bar{\sigma}$ | average volatility |
| $\varsigma$ | loading of volatility process |

**Table 2:** Variables

| Variables | Definition |
|-----------|------------|
| Endogenous | $q_t$, $\psi_t$, $\mu_t^\eta$, $\sigma_t^{q,k}$, $\sigma_t^{q,\sigma}$ |
| Secondary | $w_t^i$, $w_t^h$, $\mathbf{c}_t^i$, $\mathbf{c}_t^h$, $\iota_t^i$, $\iota_t^h$, $\mu_t^{r,i}$, $\mu_t^{r,h}$, $\mu_t^k$, $\mu_t^q$, $\mu_t^{n,i}$, $\mu_t^{n,h}$, $\mu_t^\sigma$, $\sigma_t$, $\sigma_t^{\eta,\sigma}$, $\sigma_t^{\eta,k}$, $\sigma_t^{n,i,k}$, $\sigma_t^{n,h,k}$, $\sigma_t^{n,i,\sigma}$, $\sigma_t^{n,h,\sigma}$, $\sigma_t^{\xi,i,k}$, $\sigma_t^{\xi,h,k}$, $\sigma_t^{\xi,i,\sigma}$, $\sigma_t^{\xi,h,k}$ |

Next, we focus on the variables. Model specific variables are shown in table 2. Endogenous variables are specified in the array `vars` in section *Variables* in the file `mod_BruSan.m` of the toolbox, while secondary variables are listed in `vars_`. Note that the following equations are numbered here as they are numbered in the file `mod_BruSan.m`. The superscripts $\sigma$, $\eta$, and $\xi$ are replaced with $s$, $e$, and $x$ in the code[2]. The secondary variables are defined as follows. The two state variables are

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[2]For example, $\sigma_t^{\xi,h,\sigma}$ is represented as `sigxhs`. Other variables are intuitively represented in the code. For example, $\zeta^i$ is represented as `zetai` and $\delta^h$ as `deltah`. Also note that derivatives are intuitively denoted: $q_\sigma$ and $\xi_\eta^h$ are denoted by `qz` and `veh`. Due to Matlab namespace restrictions, $\psi$ and $\phi$ are denoted by `psii` and `phii`

$$\eta_t = \mathtt{e} \tag{1}$$

$$\sigma_t = \mathtt{z} \tag{2}$$

In the code, the wealth multipliers are

$$\xi_t^i = \mathtt{vi} \tag{3}$$

$$\xi_t^h = \mathtt{vh} \tag{4}$$

Leverage was defined as

$$w_t^i = \frac{\psi_t}{\eta_t} \tag{5}$$

$$w_t^h = \frac{1 - \psi_t}{1 - \eta_t} \tag{6}$$

Consumption-to-wealth ratio is given by the first order condition:

$$c_t^i = \left(\xi_t^i\right)^{\frac{1-\zeta^i}{1-\gamma^i}} \tag{7}$$

$$c_t^h = \left(\xi_t^h\right)^{\frac{1-\zeta^h}{1-\gamma^h}} \tag{8}$$

The investment ratio is also given by its first order condition:

$$\iota_t^i = \frac{q_t - 1}{\kappa_p} \tag{9}$$

$$\iota_t^h = \frac{q_t - 1}{\kappa_p} \tag{10}$$

The functional form for $\Phi^j$ was assumed to be:

$$\Phi_t^i = log(1 + \kappa_p \iota_t^i)/\kappa_p - \delta^i \tag{11}$$

$$\Phi_t^h = log(1 + \kappa_p \iota_t^h)/\kappa_p - \delta^h \tag{12}$$

The drift of the state variable $\sigma_t$ was assumed to be

$$\mu_t^\sigma = \kappa(\sigma_t - \bar{\sigma}) \tag{13}$$

Using Ito's lemma and the process for $k_t^i$ and $k_t^h$, we get the drift of aggregate capital:

$$\mu_t^k = \psi_t \Phi_t^i + (1 - \psi_t)\Phi_t^h \tag{14}$$

From the law of motion of wealth, we have the following loadings:

$$\sigma_t^{n,i,\sigma} = w_t^i \sigma_q^{q,\sigma} \tag{15}$$

$$\sigma_t^{n,h,\sigma} = w_t^h \sigma_q^{q,\sigma} \tag{16}$$

$$\sigma_t^{n,i,k} = w_t^i (\sigma_t + \sigma_q^{q,k}) \tag{17}$$

$$\sigma_t^{n,h,k} = w_t^h (\sigma_t + \sigma_q^{q,k}) \tag{18}$$

Similarly for $\eta_t$:

$$\sigma_t^{\eta,s} = \sigma_t^{n,i,s} - \sigma_t^{q,\sigma} \tag{19}$$

$$\sigma_t^{\eta,k} = \sigma_t^{n,i,k} - (\sigma_t + \sigma_t^{q,k}) \tag{20}$$

The law of motion for $\xi_t^j$ was derived using Ito's lemma and yielded[3]

$$\sigma_t^{\xi,i,k} = \frac{\xi_\eta^i}{\xi_t^i} \sigma_t^{\eta,k} \eta_t \tag{21}$$

$$\sigma_t^{\xi,h,k} = \frac{\xi_\eta^h}{\xi_t^h} \sigma_t^{\eta,k} \eta_t \tag{22}$$

$$\sigma_t^{\xi,i,s} = \frac{\xi_\eta^i}{\xi_t^i} \sigma_t^{\eta,s} \eta_t + \frac{\xi_\sigma^i}{\xi_t^i} \varsigma \sigma_t \tag{23}$$

$$\sigma_t^{\xi,h,s} = \frac{\xi_\eta^h}{\xi_t^h} \sigma_t^{\eta,s} \eta_t + \frac{\xi_\sigma^h}{\xi_t^h} \varsigma \sigma_t \tag{24}$$

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[3]Note that $\varsigma$ is denoted as `sigs` in the code.

The same was done for

$$
\begin{aligned}
\mu_t^q = & \frac{q_\sigma}{q_t} \mu_t^\sigma \sigma_t + \frac{q_\eta}{q_t} \mu_t^\eta \eta_t + \frac{1}{2} \frac{q_{\sigma\sigma}}{q_t} \left(\varsigma\sigma_t\right)^2 \\
& + \frac{1}{2} \frac{q_{\eta\eta}}{q_t} \left[ \left((w_t^i - 1)\sigma_t^{q,\sigma}\eta_t\right)^2 + \left((w_t^i - 1)(\sigma_t + \sigma_t^{q,k})\eta_t\right)^2 \right] \\
& + \frac{q_{\sigma\eta}}{q_t} \varsigma\sigma_t \left(w_t^i - 1\right)\sigma_t^{q,\sigma}\eta_t
\end{aligned}
\tag{25}
$$

Finally, the drift of $r_t$ and $n_t^j$ using Ito's lemma:

$$
\mu_t^{r,i} = (a^i - \iota_t^i)/q_t + \Phi^i + \mu_t^q + \sigma_t\sigma_t^{q,k}
\tag{26}
$$

$$
\mu_t^{r,h} = (a^h - \iota_t^h)/q_t + \Phi^h + \mu_t^q + \sigma_t\sigma_t^{q,k}
\tag{27}
$$

$$
r_t = \mu_t^{r,i} - \gamma^i w_t^i((\sigma_t^{q,\sigma})^2 + (\sigma_t + \sigma_t^{q,k})^2) + \sigma_t^{q,\sigma}\sigma_t^{\xi,i,s} + (\sigma_t + \sigma_t^{q,k})\sigma_t^{\xi,i,k}
\tag{28}
$$

$$
\mu_t^{n,i} = r_t + w_t^i(\mu_t^{r,i} - r_t) - \mathfrak{c}_t^i
\tag{29}
$$

$$
\mu_t^{n,h} = r_t + w_t^i(\mu_t^{r,h} - r_t) - \mathfrak{c}_t^h
\tag{30}
$$

Important here is that a secondary variable cannot be used before it is specified. Once the secondary variables are defined, the endogenous variables can be specified. For each endogenous variable, an equation prefixed with `eq` is written in the code that is equal to zero. These equations are solved together to find the endogenous variable values at equilibrium. The first endogenous variable we define in the code is the drift of the state variable $\eta_t$. It was derived using Ito's lemma. The equation is written down as:

$$
\begin{aligned}
\mathtt{eqmue} = & (\mu_t^{n,i} - \mu_t^q - \mu_t^k - \sigma_t\sigma_t^{q,k} + (\sigma_t^{q,k} + \sigma_t)^2 + (\sigma_t^{q,\sigma})^2 - w^i(\sigma_t^{q,\sigma})^2 \\
& - w_t^i(\sigma_t^{q,k} + \sigma_t)^2) - \mu_t^\eta
\end{aligned}
\tag{31}
$$

The price of capital $q_t$ comes from the market clearing condition of consumption:

$$
\mathtt{eqq} = (\mathfrak{c}_t^i\eta_t + \mathfrak{c}_t^h(1 - \eta_t))q_t = (a^i - \iota_t^i)\eta_t w_t^i + (a^h - \iota_t^h)(1 - \eta_t)w_t^h
\tag{32}
$$

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$\psi_t$ solves the difference between the first order condition for $w_t^i$ and $w_t^h$:

$$
\begin{aligned}
\texttt{eqpsii} = & \mu_t^{r,i} - \mu_t^{r,h} \\
& + \gamma^h w_t^h ((\sigma_t^{q,\sigma})^2 + (\sigma_t + \sigma_t^{q,k})^2) \\
& - \gamma^i w_t^i ((\sigma_t^{q,\sigma})^2 + (\sigma_t + \sigma_t^{q,k})^2) \\
& + \sigma_t^{q,\sigma} \sigma_t^{\xi,i,s} + (\sigma_t + \sigma_t^{q,k}) \sigma_t^{\xi,i,k} \\
& - \sigma_t^{q,\sigma} \sigma_t^{\xi,h,s} + (\sigma_t + \sigma_t^{q,k}) \sigma_t^{\xi,h,k}
\end{aligned}
\tag{33}
$$

Finally, using Ito's lemma, we have:

$$
\texttt{eqsigqs} = \left( \varsigma q_\sigma \sigma_t + \sigma_t^{\eta,s} q_\eta \eta \right) - \sigma_t^{q,\sigma} q_t \tag{34}
$$

$$
\texttt{eqsigqk} = \sigma_t^{\eta,k} q_\eta \eta - \sigma_t^{q,k} q_t \tag{35}
$$

And finally, using all the optimality conditions and solved for $\mu_t^\xi$, we update the HJB in the file `HJB.m`

$$
\mu^{\xi,j} = - \left(1 - \gamma^j\right) \left( \frac{1}{1 - 1/\zeta^j} \left( \mathfrak{c}_t^j - \rho \right) + r_t - \mathfrak{c}_t^j + \frac{\gamma^j}{2} \left( w_t^j \sigma_t^{q,\sigma} \right)^2 + \frac{\gamma^j}{2} \left( w_t^j \sigma_t + w_t^j \sigma_t^{q,k} \right)^2 \right)
\tag{36}
$$