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A Proposition 1 (the Optimal Allocation)

A.1 Defining The Current Value Hamiltonian

The social planner problem is :

$$\begin{aligned}
 & \max_{\{L_{pt}, x_{it}, \tilde{x}_{it}\}} \int_0^{\infty} e^{-(\rho-gL)} L_0 u(c_t, x_{it}, \tilde{x}_{it}) dt \\
 & s.t. c_t = \frac{Y_t}{L_t} \\
 & Y_t = N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \\
 & \dot{N}_t = \frac{1}{\chi} (L_t - L_{pt}) \\
 & L_t = L_0 e^{gL^t}
 \end{aligned}$$

First, replace c_t in the problem:

$$\begin{aligned}
 & \max_{\{L_{pt}, x_{it}, \tilde{x}_{it}\}} \int_0^{\infty} e^{-(\rho-gL)} L_0 u\left(\frac{Y_t}{L_t}, x_{it}, \tilde{x}_{it}\right) dt \\
 & s.t. Y_t = N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \\
 & \dot{N}_t = \frac{1}{\chi} (L_t - L_{pt}) \\
 & L_t = L_0 e^{gL^t}
 \end{aligned}$$

Next, define Hamiltonian with state variable N_t , control variables $\{L_{pt}, x_{it}, \tilde{x}_{it}\}$ and co-state variable μ :

$$H_t(L_{pt}, x_{it}, \tilde{x}_{it}, N_t, \mu_t) = u\left(\frac{Y_t}{L_t}, x_{it}, \tilde{x}_{it}\right) + \mu_t \dot{N}_t$$

Or also:

$$H_t(L_{pt}, x_{it}, \tilde{x}_{it}, N_t, \mu_t) = u\left(\frac{Y_t}{L_t}, x_{it}, \tilde{x}_{it}\right) + \mu_t \frac{1}{\chi} (L_t - L_{pt})$$

A.2 The First-Order Conditions of the Hamiltonian

The FOC are:

$$\begin{cases} \frac{\partial H}{\partial L_{pt}} = 0 \\ \frac{\partial H}{\partial x_{it}} = 0 \\ \frac{\partial H}{\partial \tilde{x}_{it}} = 0 \\ \frac{\partial H}{\partial N_t} = (\rho - g_L)\mu_t - \dot{\mu}_t \end{cases}$$

Start with $\frac{\partial H}{\partial L_{pt}} = 0$ and recall the definition of the utility function, $u(c_t, x_{it}, \tilde{x}_{it}) = \log(c_t) - \frac{\kappa}{2} \frac{1}{N_t^2} \int_0^{N_t} x_{it}^2 di - \frac{\tilde{\kappa}}{2} \frac{1}{N_t} \int_0^{N_t} \tilde{x}_{it}^2 di$. Then:

$$\begin{aligned} \frac{\partial H}{\partial L_{pt}} &= \left(\frac{\partial u}{\partial \left(\frac{Y_t}{L_t}\right)} \right) \left(\frac{\partial \left(\frac{Y_t}{L_t}\right)}{\partial L_{pt}} \right) - \frac{\mu_t}{\chi} \\ &= \left(\frac{1}{\left(\frac{Y_t}{L_t}\right)} \right) \left(\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}}}{L_t} \frac{1}{1-\eta} L_{pt}^{\frac{\eta}{1-\eta}} \right) - \frac{\mu_t}{\chi} \\ &= \left[N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \right]^{-1} N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} \frac{1}{1-\eta} L_{pt}^{\frac{\eta}{1-\eta}} - \frac{\mu_t}{\chi} \\ &= L_{pt}^{-\frac{1}{1-\eta}} \frac{1}{1-\eta} L_{pt}^{\frac{\eta}{1-\eta}} - \frac{\mu_t}{\chi} \\ &= \frac{L_{pt}^{-1}}{1-\eta} - \frac{\mu_t}{\chi} \end{aligned}$$

Thus the FOC is:

$$\frac{L_{pt}^{-1}}{1-\eta} = \frac{\mu_t}{\chi}$$

and therefore:

$$g_\mu = -g_{L_p}$$

Next, we compute $\frac{\partial H}{\partial x_{it}} = 0$. The derivative of the Hamiltonian is given by:

$$\begin{aligned}
\frac{\partial H}{\partial x_{it}} &= \frac{\partial}{\partial x_{it}} u \left(\frac{Y_t}{L_t}, x_{it}, \tilde{x}_{it} \right) \\
&= \frac{\partial}{\partial x_{it}} u \left(\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}, x_{it}, \tilde{x}_{it} \right) \\
&= \frac{1}{\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}} \frac{L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1}}}{L_t} \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}-1} \frac{\alpha}{N_t} - \frac{\partial}{\partial x_{it}} \left(\frac{\kappa}{2N_t^2} \int_0^{N_t} x_{it}^2 di \right) \\
&= \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{-1} \frac{\alpha}{N_t} - \frac{\partial}{\partial x_{it}} \left(\frac{\kappa}{2N_t^2} \int_0^{N_t} x_{it}^2 di \right)
\end{aligned}$$

Using symmetry, $\int_0^{N_t} x_{it}^2 di = x_{it}^2 N_t$. Hence, the FOC $\frac{\partial H}{\partial x_{it}} = 0$ is:

$$\frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{-1} \alpha = \kappa x_{it}$$

Next, we compute $\frac{\partial H}{\partial \tilde{x}_{it}} = 0$. The derivative of the Hamiltonian is given by:

$$\begin{aligned}
\frac{\partial H}{\partial \tilde{x}_{it}} &= \frac{\partial}{\partial \tilde{x}_{it}} u \left(\frac{Y_t}{L_t}, x_{it}, \tilde{x}_{it} \right) \\
&= \frac{\partial}{\partial \tilde{x}_{it}} u \left(\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}, x_{it}, \tilde{x}_{it} \right) \\
&= \frac{1}{\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}} \frac{L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1}}}{L_t} \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}-1} (1-\alpha) - \tilde{\kappa} \tilde{x}_{it} \\
&= \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{-1} (1-\alpha) - \tilde{\kappa} \tilde{x}_{it}
\end{aligned}$$

Therefore the FOC $\frac{\partial H}{\partial \tilde{x}_{it}} = 0$ is:

$$\frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{-1} (1-\alpha) = \tilde{\kappa} \tilde{x}_{it}$$

Finally, the last FOC is $\frac{\partial H}{\partial N_t} = (\rho - g_L) \mu_t - \dot{\mu}_t$. The derivative of the Hamiltonian $\frac{\partial H}{\partial N_t}$

is:

$$\begin{aligned}
\frac{\partial H}{\partial N_t} &= \frac{\partial}{\partial N_t} u\left(\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}, x_{it}, \tilde{x}_{it}\right) \\
&= \frac{1}{\frac{N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}}{L_t}} \frac{L_{pt}^{\frac{1}{1-\eta}}}{L_t} \left[\frac{1}{\sigma-1} N_t^{\frac{1}{\sigma-1}-1} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{\frac{\eta}{1-\eta}} \dots \dots \right. \\
&\dots + \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{\frac{\eta}{1-\eta}-1} \left(-\frac{\alpha x_{it}}{N_t^2}\right) N_t^{\frac{1}{\sigma-1}} \left. \right] - \frac{\partial}{\partial N_t} \left(\frac{\kappa}{2} \frac{1}{N_t^2} \int_0^{N_t} x_{it}^2 di\right) - \frac{\partial}{\partial N_t} \left(\frac{\tilde{\kappa}}{2} \frac{1}{N_t} \int_0^{N_t} \tilde{x}_{it}^2 di\right) \\
&= N_t^{-1} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{-1} \left[\frac{1}{\sigma-1} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right) - \frac{\alpha x_{it}}{N_t} \right] \dots \\
&\dots - \frac{\partial}{\partial N_t} \left(\frac{\kappa}{2} \frac{1}{N_t^2} \int_0^{N_t} x_{it}^2 di\right) - \frac{\partial}{\partial N_t} \left(\frac{\tilde{\kappa}}{2} \frac{1}{N_t} \int_0^{N_t} \tilde{x}_{it}^2 di\right) \\
&= \left[\frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{-1} \right] - \frac{\partial}{\partial N_t} \left(\frac{\kappa}{2} \frac{1}{N_t^2} \int_0^{N_t} x_{it}^2 di\right) - \frac{\partial}{\partial N_t} \left(\frac{\tilde{\kappa}}{2} \frac{1}{N_t} \int_0^{N_t} \tilde{x}_{it}^2 di\right)
\end{aligned}$$

but:

$$\frac{\partial}{\partial N_t} \left(\frac{\kappa}{2} \frac{1}{N_t^2} \int_0^{N_t} x_{it}^2 di\right) = \frac{\kappa}{2} \frac{\partial}{\partial N_t} \left[x_{it}^2 \frac{1}{N_t} \right] = \frac{\kappa}{2} \left[-x_{it}^2 \frac{1}{N_t^2} \right]$$

and

$$\frac{\partial}{\partial N_t} \left(\frac{\tilde{\kappa}}{2} \frac{1}{N_t} \int_0^{N_t} \tilde{x}_{it}^2 di\right) = \frac{\tilde{\kappa}}{2} \frac{\partial}{\partial N_t} [\tilde{x}_{it}^2] = 0$$

where the last equality results from symmetry. Thus:

$$\frac{\partial H}{\partial N_t} = \left[\frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{-1} \right] + \frac{\kappa}{2} \left[x_{it}^2 \frac{1}{N_t^2} \right]$$

The FOC $\frac{\partial H}{\partial N_t} = (\rho - g_L)\mu_t - \dot{\mu}_t$ is then given by:

$$(\rho - g_L)\mu_t - \dot{\mu}_t = \frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it}\right)^{-1} + \frac{\kappa}{2} x_{it}^2 \frac{1}{N_t^2}$$

A.3 Solving the Optimal Allocation

The 4 FOCs are:

$$\left\{ \begin{array}{l} \frac{L_{pt}^{-1}}{1-\eta} = \frac{\mu_t}{\chi} \quad (A) \\ \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} \alpha = \kappa x_{it} \quad (B) \\ \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} (1-\alpha) = \tilde{\kappa}\tilde{x}_{it} \quad (C) \\ (\rho - g_L)\mu_t - \dot{\mu}_t = \frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} + \frac{\kappa}{2} x_{it}^2 \frac{1}{N_t^2} \quad (D) \end{array} \right.$$

Solving for \tilde{x}_{it} and x_{it} (eqs. 24 and 25).

Divide (B) by (C):

$$\frac{\alpha}{(1-\alpha)} = \frac{\kappa}{\tilde{\kappa}} \left(\frac{x_{it}}{\tilde{x}_{it}} \right)$$

so

$$x_{it} = \frac{\alpha}{(1-\alpha)} \left(\frac{\tilde{\kappa}}{\kappa} \right) \tilde{x}_{it}$$

Replace in (C):

$$\begin{aligned} \frac{\eta}{1-\eta} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} (1-\alpha) &= \tilde{\kappa}\tilde{x}_{it} \\ \Rightarrow \frac{\eta}{1-\eta} \left(\frac{\alpha \frac{\alpha}{(1-\alpha)} \left(\frac{\tilde{\kappa}}{\kappa} \right) \tilde{x}_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} (1-\alpha) &= \tilde{\kappa}\tilde{x}_{it} \end{aligned}$$

Solve for \tilde{x}_{it} :

$$\tilde{x}_{it}^2 = \left(\frac{\eta}{1-\eta} \frac{1}{\tilde{\kappa}} \right) \left[\frac{(1-\alpha)}{\frac{\alpha \frac{\alpha}{(1-\alpha)} \left(\frac{\tilde{\kappa}}{\kappa} \right)}{N_t} + (1-\alpha)} \right]$$

and $\lim_{N_t \rightarrow \infty} \left[\frac{(1-\alpha)}{\frac{\alpha \frac{\alpha}{(1-\alpha)} \left(\frac{\tilde{\kappa}}{\kappa} \right)}{N_t} + (1-\alpha)} \right] = 1$. Hence the optimal values for \tilde{x}_{it} and x_{it} as N_t

grows large are:

$$\tilde{x}_{it}^{sp} = \left(\frac{\eta}{1-\eta} \frac{1}{\tilde{\kappa}} \right)^{\frac{1}{2}} \quad (24)$$

$$x_{it}^{sp} = \frac{\alpha}{(1-\alpha)} \left(\frac{\tilde{\kappa}}{\kappa} \right) \left(\frac{\eta}{1-\eta} \frac{1}{\tilde{\kappa}} \right)^{\frac{1}{2}}. \quad (25)$$

Solving for L_{it}^{sp} (eq. 26).

Using (A):

$$g_{L_{pt}} = -\frac{\dot{\mu}}{\mu}$$

Using this and (A) in (D):

$$(\rho - g_L) - \frac{\dot{\mu}_t}{\mu} = \frac{1}{\mu} \left[\frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} \frac{\kappa}{2} x_{it}^2 \frac{1}{N_t^2} \right]$$

$$\Rightarrow (\rho - g_L) + g_{L_{pt}} = \frac{L_{pt}(1-\eta)}{\chi} \left[\frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} + \frac{\kappa}{2} x_{it}^2 \frac{1}{N_t^2} \right]$$

but $g_L = g_{L_p}$ ¹¹ therefore:

$$\Rightarrow \rho = \frac{L_{pt}(1-\eta)}{\chi} \left[\frac{1}{(\sigma-1)N_t} - \frac{\alpha x_{it}}{N_t^2} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)^{-1} + \frac{\kappa}{2} x_{it}^2 \frac{1}{N_t^2} \right]$$

use the approximation $\frac{1}{N_t^2} \simeq 0$ when N_t grows large, then:

$$L_{pt} = \frac{\rho\chi(\sigma-1)}{1-\eta} N_t$$

and also:

¹¹Since $\dot{N}_t = \frac{1}{\chi} L_{et}$, then because of BGP $g_N = \frac{1}{\chi} \frac{L_{et}}{N_t}$, and therefore $g_{L_{et}} = g_N$ since the RHS must be constant in time. On the other hand, $\dot{N}_t = \frac{1}{\chi} (L_t - L_{pt})$, hence $g_N = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$, and since the RHS must be constant then $g_L = g_{L_{pt}} = g_N$.

$$g_N = g_{L_p}$$

Now use $L_{it} = \frac{L_{pt}}{N_t}$ and therefore:

$$L_{it}^{sp} = \frac{\rho\chi(\sigma-1)}{1-\eta} := \nu_{sp} \quad (26)$$

Solving for N_t^{sp} (eq. 27).

From the definition of the innovation process, $\frac{\dot{N}_t}{N_t} = \frac{1}{\chi}(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t})$ and therefore

$$\frac{L_t}{N_t} = \frac{\dot{N}_t}{N_t}\chi + \frac{L_{pt}}{N_t}$$

but from above $\frac{\dot{N}_t}{N_t} = g_N = g_{L_p}$ and $g_{L_p} = g_L$ therefore $\frac{\dot{N}_t}{N_t} = g_L$ and we get:

$$\frac{L_t}{N_t} = g_L\chi + \frac{L_{pt}}{N_t}$$

and from above we know that when N_t grows large, then $L_{pt} = \frac{\rho\chi(\sigma-1)}{1-\eta}N_t$, or also: $\frac{L_{pt}}{N_t} = \frac{\rho\chi(\sigma-1)}{1-\eta}$. Thus when N_t grows large we have:

$$\frac{L_t}{N_t} = g_L\chi + \frac{\rho\chi(\sigma-1)}{1-\eta}$$

or also:

$$N_t = L_t \left[\frac{1}{g_L\chi + \frac{\rho\chi(\sigma-1)}{1-\eta}} \right]$$

but $\frac{\rho\chi(\sigma-1)}{1-\eta} = \nu_{sp}$ and therefore we finally get:

$$N_t^{sp} = \frac{L_t}{\chi g_L + \nu_{sp}} = \psi_{sp} L_t \quad (27)$$

Solving for L_{pt}^{sp} (eq. 28).

Use $L_{pt} = \frac{\rho\chi(\sigma-1)}{1-\eta}N_t$ and equation (25) $N_t^{sp} = \frac{L_t}{\chi g_L + \nu_{sp}} = \psi_{sp} L_t$ to get :

$$\begin{aligned}
L_{pt}^{sp} &= \frac{\rho\chi(\sigma-1)}{1-\eta} N_t \\
&= \frac{\rho\chi(\sigma-1)}{1-\eta} \psi_{sp} L_t \\
&= \nu_{sp} \psi_{sp} L_t
\end{aligned} \tag{28}$$

Solving for Y_t^{sp} (eq. 29).

Aggregate production is given by $Y_t = N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}$. Use the optimal expression for N_t^{sp} , L_{pt}^{sp} , \tilde{x}_{sp} and x_{sp} to get:

$$\begin{aligned}
Y_t &= N_t^{\frac{1}{\sigma-1}} \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \\
&= (\psi_{sp} L_t)^{\frac{1}{\sigma-1}} \left(\alpha \frac{\alpha}{N_t} \left(\frac{\tilde{\kappa}}{\kappa} \right) \tilde{x}_{sp} + (1-\alpha) \tilde{x}_{sp} \right)^{\frac{\eta}{1-\eta}} (\nu_{sp} \psi_{sp} L_t)^{\frac{1}{1-\eta}} \\
&= [v_{sp}]^{\frac{1}{1-\eta}} \left(\alpha \frac{\alpha}{N_t} \left(\frac{\tilde{\kappa}}{\kappa} \right) \tilde{x}_{sp} + (1-\alpha) \tilde{x}_{sp} \right)^{\frac{\eta}{1-\eta}} (\psi_{sp} L_t)^{\frac{1}{1-\eta} + \frac{1}{\sigma-1}}
\end{aligned}$$

Assuming $\alpha \frac{\alpha}{N_t} \left(\frac{\tilde{\kappa}}{\kappa} \right) \tilde{x}_{sp} \simeq 0$ when N_t is high, then:

$$Y_t = [v_{sp}(1-\alpha)^{\eta} \tilde{x}_{sp}^{\eta}]^{\frac{1}{1-\eta}} (\psi_{sp} L_t)^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}} \tag{29}$$

Solving for c_t^{sp} (eq. 30).

By definition, consumption per capita is:

$$\begin{aligned}
c_t &= \frac{Y_t}{L_t} \\
&= \frac{[v_{sp}(1-\alpha)^{\eta} \tilde{x}_{sp}^{\eta}]^{\frac{1}{1-\eta}} (\psi_{sp} L_t)^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}}}{L_t} \\
&= [v_{sp}(1-\alpha)^{\eta} \tilde{x}_{sp}^{\eta}]^{\frac{1}{1-\eta}} (\psi_{sp})^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}} L_t^{\frac{1}{\sigma-1} + \frac{\eta}{1-\eta}}
\end{aligned} \tag{30}$$

Solving for g_c^{sp} (eq. 31).

Using the definition of consumption per capita growth and equation 30 yields:

$$\begin{aligned}
g_c^{sp} &= g_{Y/L}^{sp} \\
&= \left(\frac{1}{\sigma - 1} + \frac{\eta}{1 - \eta} \right) g_L
\end{aligned} \tag{31}$$

Solving for D_{it}^{sp} (eq. 32).

From equation 19, $Y_t = N_t^{\frac{1}{\sigma-1}} D_{it}^\eta L_{pt}$. Plugging the previous results yields:

$$\begin{aligned}
D_{it}^{sp} &= \left(\frac{Y_t}{N_t^{\frac{1}{\sigma-1}} L_{pt}} \right)^{\frac{1}{\eta}} \\
&= \left(\frac{[v_{sp}(1 - \alpha)^\eta \tilde{x}_{sp}^\eta]^\frac{1}{1-\eta} (\psi_{sp} L_t)^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}}}{(\psi_{sp} L_t)^{\frac{1}{\sigma-1}} \nu_{sp} \psi_{sp} L_t} \right)^{\frac{1}{\eta}} \\
&= \left([(1 - \alpha) \tilde{x}_{sp}]^{\frac{\eta}{1-\eta}} (v_{sp} \psi_{sp} L_t)^{\frac{\eta}{1-\eta}} \right)^{\frac{1}{\eta}} \\
&= [(1 - \alpha) \tilde{x}_{sp} v_{sp} \psi_{sp} L_t]^{\frac{1}{1-\eta}}
\end{aligned} \tag{32}$$

Solving for D_t^{sp} (eq. 33).

By definition:

$$\begin{aligned}
D_t^{sp} &= N_t^{sp} D_{it}^{sp} \\
&= (\psi_{sp} L_t) \left([(1 - \alpha) \tilde{x}_{sp} v_{sp} \psi_{sp} L_t]^{\frac{1}{1-\eta}} \right) \\
&= [(1 - \alpha) \tilde{x}_{sp} v_{sp}]^{\frac{1}{1-\eta}} (\psi_{sp} L_t)^{\frac{1}{1-\eta} + 1}
\end{aligned} \tag{33}$$

Solving for Y_{it}^{sp} (eq. 34).

Using equation 18 and the results for L_{pt} and D_{it} :

$$\begin{aligned}
Y_{it}^{sp} &= D_{it}^\eta \frac{L_{pt}}{N_t} \\
&= [(1 - \alpha) \tilde{x}_{sp} v_{sp} \psi_{sp} L_t]^{\frac{\eta}{1-\eta}} (\nu_{sp}) \\
&= \left[(1 - \alpha)^\eta \tilde{x}_{sp}^\eta v_{sp} \right]^{\frac{1}{1-\eta}} (\psi_{sp} L_t)^{\frac{\eta}{1-\eta}}
\end{aligned} \tag{34}$$

Solving for Y_0^{sp} (eq. 35).

By definition $U_0 = \int_0^\infty e^{-(\rho-g_L)t} L_0 u(c_t, x_{it}, \tilde{x}_{it}) dt$. Then:

$$\begin{aligned} U_0 &= \int_0^\infty e^{-(\rho-g_L)t} L_0 u(c_t, x_{it}, \tilde{x}_{it}) dt \\ &= L_0 \int_0^\infty e^{-(\tilde{\rho})t} \left[\log([v_{sp}(1-\alpha)^\eta \tilde{x}_{sp}^\eta]^{\frac{1}{1-\eta}} (\psi_{sp})^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}} L_t^{\frac{1}{\sigma-1} + \frac{\eta}{1-\eta}}) \dots \right. \\ &\quad \left. \dots - \frac{\kappa}{2N_t} x_{sp}^2 - \frac{\tilde{\kappa}}{2} \tilde{x}_{sp}^2 \right] \end{aligned}$$

Next, assume $\frac{\kappa}{2N_t} x_{sp}^2 \simeq 0$ and use $c_0 = [v_{sp}(1-\alpha)^\eta \tilde{x}_{sp}^\eta]^{\frac{1}{1-\eta}} (\psi_{sp})^{\frac{1}{\sigma-1} + \frac{1}{1-\eta}} L_0^{\frac{1}{\sigma-1} + \frac{\eta}{1-\eta}}$.

Then:

$$\begin{aligned} U_0 &= L_0 \int_0^\infty e^{-(\tilde{\rho})t} \left[\log(c_0) + \left(\frac{1}{\sigma-1} + \frac{\eta}{1-\eta} \right) g_L t \right] dt + \dots \\ &\quad - L_0 \frac{\tilde{\kappa}}{2} \tilde{x}_{sp}^2 \int_0^\infty e^{-(\tilde{\rho})t} dt \\ &= L_0 \frac{1}{\tilde{\rho}} \left(\log(c_0) - \frac{\tilde{\kappa}}{2} \tilde{x}_{sp}^2 \right) + L_0 \left(\frac{1}{\sigma-1} + \frac{\eta}{1-\eta} \right) g_L \int_0^\infty e^{-(\tilde{\rho})t} t dt \\ &= L_0 \frac{1}{\tilde{\rho}} \left(\log(c_0) - \frac{\tilde{\kappa}}{2} \tilde{x}_{sp}^2 \right) + L_0 \left(\frac{1}{\sigma-1} + \frac{\eta}{1-\eta} \right) g_L \frac{1}{\tilde{\rho}^2} \\ &= L_0 \frac{1}{\tilde{\rho}} \left(\log(c_0) - \frac{\tilde{\kappa}}{2} \tilde{x}_{sp}^2 + \frac{g_c}{\tilde{\rho}} \right) \tag{35} \end{aligned}$$

B Competitive Equilibrium When Firms Own Data

B.1 Household Problem

The problem is defined by:

$$\begin{aligned} U_0 &= \max_{\{c_{it}\}} \int_0^\infty e^{-(\tilde{\rho})t} L_0 u(c_t, x_{it}, \tilde{x}_{it}) dt \\ s.t. \quad c_t &= \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di \right)^{\frac{\sigma}{\sigma-1}} \\ \dot{a}_t &= (r_t - g_L) a_t + w_t - \int_0^{N_t} p_{it} c_{it} di \end{aligned}$$

Define the current value Hamiltonian with state variable a_t , control variable c_{it} and co-state variable μ_t :

$$H(a_t, c_{it}, \mu_t) = u\left(\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{\sigma}{\sigma-1}}, x_{it}, \tilde{x}_{it}\right) + \mu_t \left[(r_t - g_L)a_t + w_t - \int_0^{N_t} p_{it} c_{it} di \right]$$

The FOCs are:

$$\begin{cases} \frac{\partial H}{\partial c_{it}} = 0 \\ \frac{\partial H}{\partial a_t} = \tilde{\rho}\mu_t - \dot{\mu}_t \end{cases}$$

First, start with $\frac{\partial H}{\partial c_{it}} = 0$:

$$\frac{1}{\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{\sigma}{\sigma-1}}} \frac{\sigma}{\sigma-1} \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} c_{it}^{\frac{-1}{\sigma}} - \mu_t p_{it} = 0$$

$$\Rightarrow c_t^{\frac{1-\sigma}{\sigma}} c_{it}^{\frac{-1}{\sigma}} - \mu_t p_{it} = 0$$

$$\Rightarrow c_t^{\sigma-1} c_{it} = (\mu_t p_{it})^{-\sigma}$$

$$\Rightarrow (c_t^{\sigma-1} c_{it})^{\frac{\sigma-1}{\sigma}} = (\mu_t p_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_{it}^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} (p_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} \int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_t^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{\sigma-1}{\sigma} + \frac{(1-\sigma)^2}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow \mu_t = \frac{1}{c_t \left(\int_0^{N_t} p_{it}^{1-\sigma} di \right)^{\frac{1}{1-\sigma}}}$$

Next, use the fact that the price of c_t is normalized to 1, i.e., $P_t = \left(\int_0^{N_t} p_{it}^{1-\sigma} di \right)^{\frac{1}{1-\sigma}} =$

1. Next, plug the expression for μ_t in the FOC and this gives equation (39):

$$c_{it} = c_t P_{it}^{-\sigma}$$

Next, compute the FOC $\frac{\partial H}{\partial a_t} = \tilde{\rho}\mu_t - \dot{\mu}_t$:

$$\mu_t(r_t - g_L) = \tilde{\rho}\mu_t - \dot{\mu}_t$$

$$\Rightarrow (r_t - g_L) = \tilde{\rho} - \frac{\dot{\mu}_t}{\mu_t}$$

But $\mu_t = c_t^{-1}$ then $\frac{\dot{\mu}_t}{\mu_t} = -g_c$. Thus:

$$\Rightarrow (r_t - g_L) = \tilde{\rho} + g_c$$

B.2 Firm Problem

The firm problem is:

$$r_t V_{it} = \max_{\{L_{it}, D_{bit}, x_{it}, \tilde{x}_{it}\}} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} - p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it} - \delta(\tilde{x}_{it}) V_{it} + \dot{V}_{it}$$

$$s.t. Y_{it} = D_{it}^\eta L_{it}$$

$$D_{it} = \alpha x_{it} Y_{it} + (1 - \alpha) D_{bit}$$

$$p_{sit} = \lambda_{DI} N_t^{-\frac{1}{\epsilon}} \left(\frac{B_t}{\tilde{x}_{it} Y_{it}} \right)^{\frac{1}{\epsilon}}$$

$$x_{it} \in [0; 1]$$

$$\tilde{x}_{it} \in [0; 1]$$

taking as given λ_{DI} , B_t , N_t , p_{bt} and Y_t . To solve this problem, write the Lagrangean:

$$\begin{aligned}\mathbb{L} = & (Y_t)^{\frac{1}{\sigma}} Y_{it}^{1-\frac{1}{\sigma}} - w_t L_{it} - p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it} - \delta(\tilde{x}_{it}) V_{it} + \dot{V}_{it} \dots \\ & + \mu_{x0}(x_{it}) + \mu_{\tilde{x}0}(\tilde{x}_{it}) + \mu_{\tilde{x}1}(1 - \tilde{x}_{it}) + \mu_{x1}(1 - x_{it}) + \mu_{0d} D_{bit}\end{aligned}$$

Simplify using the constraints:

$$\begin{aligned}\mathbb{L} = & (Y_t)^{\frac{1}{\sigma}} Y_{it}^{1-\frac{1}{\sigma}} - w_t L_{it} - p_{bt} D_{bit} + \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (\tilde{x}_{it} Y_{it})^{1-\frac{1}{\epsilon}} - \delta(\tilde{x}_{it}) V_{it} + \dot{V}_{it} \dots \\ & + \mu_{x0}(x_{it}) + \mu_{\tilde{x}0}(\tilde{x}_{it}) + \mu_{\tilde{x}1}(1 - \tilde{x}_{it}) + \mu_{x1}(1 - x_{it})\end{aligned}$$

Now take the FOCs.

1) Start with the FOC w.r.t. to L_{it} :

$$\begin{aligned}\frac{\partial \mathbb{L}}{\partial L_{it}} &= 0 \\ \Leftrightarrow (1 - \frac{1}{\sigma}) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} \frac{\partial Y_{it}}{\partial L_{it}} - w_t + (1 - \frac{1}{\epsilon}) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (\tilde{x}_{it})^{1-\frac{1}{\epsilon}} \frac{\partial Y_{it}}{\partial L_{it}} &= 0 \\ \frac{\partial Y_{it}}{\partial L_{it}} \left[(1 - \frac{1}{\sigma}) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + (1 - \frac{1}{\epsilon}) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] &= w_t\end{aligned}$$

And using $Y_{it} = D_{it}^{\eta} L_{it}$ and assuming D_{it} depends on L_{it} , then by implicit derivation:

$$\begin{aligned}\frac{\partial Y_{it}}{\partial L_{it}} &= \eta D_{it}^{\eta-1} L_{it} \frac{\partial D_{it}}{\partial L_{it}} + D_{it}^{\eta} \\ \Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} &= \eta \frac{Y_{it}}{D_{it}} \frac{\partial D_{it}}{\partial L_{it}} + \frac{Y_{it}}{L_{it}}\end{aligned}$$

Next, compute $\frac{\partial D_{it}}{\partial L_{it}}$ using $D_{it} = \alpha x_{it} Y_{it} + (1 - \alpha) D_{bit}$:

$$\frac{\partial D_{it}}{\partial L_{it}} = \alpha x_{it} \frac{\partial Y_{it}}{\partial L_{it}}$$

Substituting above:

$$\Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} = \eta \frac{Y_{it}}{D_{it}} \alpha x_{it} \frac{\partial Y_{it}}{\partial L_{it}} + \frac{Y_{it}}{L_{it}}$$

$$\Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} = \frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}}$$

The FOC for L_{it} is then:

$$\frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t$$

2) Compute the FOC w.r.t. to D_{bit} :

$$\begin{aligned} \frac{\partial \mathbb{L}}{\partial D_{bit}} &= 0 \\ \Leftrightarrow \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \frac{\partial Y_{it}}{\partial D_{bit}} - w_t + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (\tilde{x}_{it})^{1-\frac{1}{\epsilon}} \frac{\partial Y_{it}}{\partial D_{bit}} - p_{bt} + \mu_{0d} &= 0 \\ \frac{\partial Y_{it}}{\partial D_{bit}} \left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] &= p_{bt} \end{aligned}$$

And using $Y_{it} = D_{it}^\eta L_{it}$ we have:

$$\frac{\partial Y_{it}}{\partial D_{bit}} = \eta \frac{Y_{it}}{D_{it}} \frac{\partial D_{it}}{\partial D_{bit}}$$

and using $D_{it} = \alpha x_{it} Y_{it} + (1 - \alpha) D_{bit}$:

$$\frac{\partial D_{it}}{\partial D_{bit}} = \alpha x_{it} \frac{\partial Y_{it}}{\partial D_{bit}} + (1 - \alpha)$$

Substituting above:

$$\begin{aligned} \frac{\partial Y_{it}}{\partial D_{bit}} &= \eta \frac{Y_{it}}{D_{it}} \left(\alpha x_{it} \frac{\partial Y_{it}}{\partial D_{bit}} + (1 - \alpha) \right) \\ \Rightarrow \frac{\partial Y_{it}}{\partial D_{bit}} &= \frac{(1 - \alpha) \eta \frac{Y_{it}}{D_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \end{aligned}$$

Then the FOC for D_{bit} is

$$\frac{(1-\alpha)\eta\frac{Y_{it}}{D_{it}}}{1-\eta\frac{Y_{it}}{D_{it}}\alpha x_{it}} \left[\left(1-\frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1-\frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = p_{bt}$$

3) Compute the FOC w.r.t. to x_{it} :

$$\begin{aligned} \frac{\partial \mathbb{L}}{\partial x_{it}} &= 0 \\ \Leftrightarrow \left(1-\frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \frac{\partial Y_{it}}{\partial x_{it}} + \left(1-\frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (\tilde{x}_{it})^{1-\frac{1}{\epsilon}} \frac{\partial Y_{it}}{\partial x_{it}} + \mu_{x0} - \mu_{x1} &= 0 \\ \frac{\partial Y_{it}}{\partial x_{it}} \left[\left(1-\frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1-\frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] &= -\mu_{x0} + \mu_{x1} \end{aligned}$$

Now to compute $\frac{\partial Y_{it}}{\partial x_{it}}$ use $Y_{it} = D_{it}^{\eta} L_{it}$:

$$\frac{\partial Y_{it}}{\partial x_{it}} = \eta \frac{Y_{it}}{D_{it}} \frac{\partial D_{it}}{\partial x_{it}}$$

and using $D_{it} = \alpha x_{it} Y_{it} + (1-\alpha) D_{bit}$ and implicit derivation:

$$\frac{\partial D_{it}}{\partial x_{it}} = \alpha x_{it} \frac{\partial Y_{it}}{\partial x_{it}} + Y_{it} \alpha$$

Thus:

$$\begin{aligned} \Rightarrow \frac{\partial Y_{it}}{\partial x_{it}} &= \eta \frac{Y_{it}}{D_{it}} \left[\alpha x_{it} \frac{\partial Y_{it}}{\partial x_{it}} + Y_{it} \alpha \right] \\ \Rightarrow \frac{\partial Y_{it}}{\partial x_{it}} &= \frac{\eta \frac{Y_{it}}{D_{it}} Y_{it} \alpha}{1 - \alpha x_{it} \eta \frac{Y_{it}}{D_{it}}} \end{aligned}$$

Then the FOC w.r.t. to x_{it} is:

$$\frac{\eta \frac{Y_{it}}{D_{it}} Y_{it} \alpha}{1 - \alpha x_{it} \eta \frac{Y_{it}}{D_{it}}} \left[\left(1-\frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1-\frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = -\mu_{x0} + \mu_{x1}$$

Now, note that the LHS is > 0 , then:

$$\mu_{x1} > \mu_{x0} \geq 0$$

$$\Rightarrow \mu_{x1} > 0$$

$$\Rightarrow x_{it} = 1 \tag{A.1}$$

4) Compute the FOC w.r.t. to \tilde{x}_{it} :

$$\begin{aligned} \frac{\partial \mathbb{L}}{\partial \tilde{x}_{it}} &= 0 \\ \Leftrightarrow \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (Y_{it})^{1-\frac{1}{\epsilon}} \tilde{x}_{it}^{-\frac{1}{\epsilon}} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it}) V_{it} + \mu_{\tilde{x}0} - \mu_{\tilde{x}1} &= 0 \\ \lambda_{DI} N_t^{-\frac{1}{\epsilon}} (B_t)^{\frac{1}{\epsilon}} (Y_{it})^{1-\frac{1}{\epsilon}} \tilde{x}_{it}^{-\frac{1}{\epsilon}} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it}) V_{it} - \mu_{\tilde{x}0} + \mu_{\tilde{x}1} &= 0 \end{aligned}$$

Or also:

$$p_{sit} Y_{it} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it}) V_{it} = -\mu_{\tilde{x}0} + \mu_{\tilde{x}1}$$

Finally, the 4 FOCs of the firm problem are:

$$\left\{ \begin{array}{l} \frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t \quad (A) \\ \frac{(1-\alpha)\eta \frac{Y_{it}}{D_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = p_{bt} \quad (B) \\ x_{it} = 1 \quad (C) \\ p_{sit} Y_{it} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it}) V_{it} + \mu_{\tilde{x}0} - \mu_{\tilde{x}1} = 0 \quad (D) \end{array} \right.$$

5) Solution in terms of equilibrium aggregates

Divide (A) by (B):

$$\begin{aligned} \left[\frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \right] \left[\frac{(1 - \alpha) \eta \frac{Y_{it}}{D_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \right]^{-1} &= \frac{w_t}{p_{bt}} \\ \Rightarrow \frac{D_{it}}{(1 - \alpha) \eta L_{it}} &= \frac{w_t}{p_{bt}} \end{aligned}$$

Hence:

$$\Rightarrow D_{it} = \frac{w_t}{p_{bt}} (1 - \alpha) \eta L_{it} \quad (\text{A.2})$$

Next, substitute in (A) and use (C):

$$\frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha} \left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t$$

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t \frac{L_{it}}{Y_{it}} \left(1 - \eta \frac{Y_{it}}{D_{it}} \alpha\right)$$

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t \frac{L_{it}}{Y_{it}} - w_t \eta \frac{L_{it}}{D_{it}} \alpha$$

But from above $D_{it} = \frac{w_t}{p_{bt}} (1 - \alpha) \eta L_{it}$, therefore substitute in RHS:

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} \right] = w_t \frac{L_{it}}{Y_{it}} - \alpha \left(\frac{p_{bt}}{1 - \alpha}\right)$$

Then:

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} + \alpha \left(\frac{p_{bt}}{1 - \alpha}\right) \right] \frac{Y_{it}}{L_{it}} = w_t \quad (\text{A.3})$$

Next, we need to compute $\frac{Y_{it}}{L_{it}}$ using: $Y_{it} = D_{it}^\eta L_{it}$:

$$\frac{Y_{it}}{L_{it}} = D_{it}^\eta = \left[\frac{w_t}{p_{bt}} (1 - \alpha) \eta \right]^\eta L_{it}^\eta \quad (\text{A.4})$$

B.3 Data Intermediary Problem

The problem faced by the data intermediary is :

$$\begin{aligned} \max_{p_{bt}, D_{sit}} \quad & p_{bt} \int_0^{N_t} D_{bit} di - \int_0^{N_t} p_{sit} D_{sit} di \\ \text{s.t.} \quad & D_{bit} \leq B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \\ & p_{bt} \leq p_{bt}^* \end{aligned}$$

where p_{sit} , i.e. the purchase price of data is taken as given.

B.3.1 The downward-sloping demand curve: data intermediary cost minimization

To compute the demand curve of the data intermediary, we solve the following cost minimization problem:

$$\begin{aligned} \min_{D_{sit}} \quad & \int_0^{N_t} p_{sit} D_{sit} di \\ \text{s.t.} \quad & D_{bit} \leq B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \end{aligned}$$

The Lagrangean is given by:

$$\mathbb{L} = \int_0^{N_t} p_{sit} D_{sit} di + \lambda_{DI} \left[D_{bit} - \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \right]$$

By symmetry:

$$\mathbb{L} = \int_0^{N_t} p_{sit} D_{sit} di + \lambda_{DI} \left[D_{bit} - \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \right]$$

Taking FOC yields:

$$\begin{aligned} \frac{\partial \mathbb{L}}{\partial D_{sit}} &= 0 \\ \Leftrightarrow p_{sit} - \lambda_{DI} \frac{\epsilon}{\epsilon-1} \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}-1} \frac{\epsilon-1}{\epsilon} D_{sit}^{\frac{\epsilon-1}{\epsilon}-1} &= 0 \\ p_{sit} - \lambda_{DI} B_t^{\frac{1}{\epsilon}} D_{sit}^{\frac{\epsilon-1}{\epsilon}-1} N_t^{-\frac{1}{\epsilon}} &= 0 \\ \lambda_{DI} \left(\frac{B_t}{D_{sit}} \right)^{\frac{1}{\epsilon}} N_t^{-\frac{1}{\epsilon}} &= p_{sit} \end{aligned}$$

We get the following demand curve:

$$p_{sit} = \lambda_{DI} N_t^{-\frac{1}{\epsilon}} \left(\frac{B_t}{D_{sit}} \right)^{\frac{1}{\epsilon}}$$

which is equation (44), which is used as a constraint in the firm problem. Next, by symmetry

$$B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} (D_{sit})^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} = N_t^{\frac{1}{\epsilon} \frac{\epsilon}{\epsilon-1} + \frac{\epsilon}{\epsilon-1}} D_{sit} = N_t D_{sit} \quad (\text{A.5})$$

Substituting in equation (44)

$$p_{sit} = \lambda_{DI} N_t^{-\frac{1}{\epsilon}} \left(\frac{N_t D_{sit}}{D_{sit}} \right)^{\frac{1}{\epsilon}}$$

$$\Rightarrow p_{sit} = \lambda_{DI}$$

B.3.2 The zero profit condition

The objective function is increasing in p_{bt} . It is clear that $p_{bt} = p_{bt}^*$, which is the price given by the zero profit condition. Use the zero profit condition:

$$\begin{aligned} \Pi &= 0 \\ p_{bt} \int_0^{N_t} D_{bit} di - \int_0^{N_t} p_{sit} D_{sit} di &= 0 \\ p_{bt} \int_0^{N_t} B_t di - \int_0^{N_t} p_{sit} D_{sit} di &= 0 \\ p_{bt} B_t N_t - p_{sit} D_{sit} N_t &= 0 \end{aligned}$$

Thus we get:

$$p_{bt} = \frac{p_{sit} D_{sit}}{B_t}$$

Next, from the demand curve, we have $\lambda_{DI} \left(\frac{B_t}{D_{sit}} \right)^{\frac{1}{\epsilon}} N_t^{-\frac{1}{\epsilon}} = p_{sit}$ and solving for B_t we get $B_t = \left(\frac{p_{sit}}{\lambda_{DI}} \right)^{\epsilon} N_t D_{sit} = N_t D_{sit}$ where the last equation comes from the fact that $p_{sit} = \lambda_{DI}$. Finally:

$$\begin{aligned}
p_{bt} &= \frac{p_{sit} D_{sit}}{N_t D_{sit}} \\
\Rightarrow p_{bt} &= \frac{p_{sit}}{N_t} \tag{A.6}
\end{aligned}$$

B.4 Free Entry and the Creation of New Varieties

The free entry condition is given by:

$$\chi w_t = V_{it} + \frac{\int_0^{N_t} \delta(\tilde{x}_{it}) V_{it} di}{\dot{N}_t}$$

by symmetry:

$$\begin{aligned}
\chi w_t &= V_{it} + \frac{\delta(\tilde{x}_{it}) V_{it}}{\frac{\dot{N}_t}{N_t}} \\
\chi w_t &= V_{it} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L} \right)
\end{aligned}$$

B.5 Equilibrium when Firms Own Data

B.5.1 Solve for p_{bt} , Y_{it} , D_{sit} , D_{bit} , w_t as a function of \tilde{x}_{it} and aggregates

Take the FOC w.r.t. to L_i in the firm problem, i.e. A.3:

$$\left[\left(1 - \frac{1}{\sigma} \right) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon} \right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} + \alpha \left(\frac{p_{bt}}{1-\alpha} \right) \right] \left[\frac{w_t}{p_{bt}} (1-\alpha) \eta \right]^\eta L_{it}^\eta = w_t$$

Now, from the date intermediary problem, i.e. A.6, $p_{bt} = \frac{p_{sit}}{N_t} = \lambda_{DI} N_t^{-\frac{1}{\epsilon}-1} \left(\frac{B_t}{\tilde{x}_{it} Y_{it}} \right)^{\frac{1}{\epsilon}}$.

Substitute above:

$$\left[\left(1 - \frac{1}{\sigma} \right) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon} \right) \tilde{x}_{it} p_{bt} N_t + \alpha \left(\frac{p_{bt}}{1-\alpha} \right) \right] \left[\frac{w_t}{p_{bt}} (1-\alpha) \eta \right]^\eta L_{it}^\eta = w_t$$

Next, from A.2, $D_{it} = \frac{w_t}{p_{bt}}(1 - \alpha)\eta L_{it}$, then

$$p_{bt} = \frac{w_t}{D_{it}}(1 - \alpha)\eta L_{it}$$

Use (A.4) to get $\frac{Y_{it}}{D_{it}} = D_{it}^{\eta-1} L_{it} = \left[\frac{w_t}{p_{bt}}(1 - \alpha)\eta \right]^{\eta-1} L_{it}^\eta$. Substitute:

$$\begin{aligned} p_{bt} &= \frac{w_t}{D_{it}}(1 - \alpha)\eta L_{it} \\ &= \left[\frac{w_t}{p_{bt}}(1 - \alpha)\eta \right]^{\eta-1} L_{it}^\eta w_t(1 - \alpha)\eta \left(\frac{Y_{it}}{L_{it}} \right)^{-1} \end{aligned}$$

Next, from equation (13), by symmetry, we know that $L_{it} = \frac{L_{pt}}{N_t}$. Moreover, by symmetry, $c_t = N_t^{\frac{\sigma}{\sigma-1}} c_{it} = N_t^{\frac{\sigma}{\sigma-1}} \frac{Y_{it}}{L_{it}}$ and therefore $Y_t = c_t L_t = N_t^{\frac{\sigma}{\sigma-1}} \frac{Y_{it}}{L_{it}} L_t = N_t^{\frac{\sigma}{\sigma-1}} Y_{it}$.

Thus $\frac{Y_{it}}{L_{it}} = \left(\frac{Y_t}{N_t^{\frac{\sigma}{\sigma-1}} L_{pt}} \right) = N_t^{-\frac{1}{\sigma-1}} \frac{Y_t}{L_{pt}}$. Substituting above:

$$\begin{aligned} p_{bt} &= \left[\frac{w_t}{p_{bt}}(1 - \alpha)\eta \right]^{\eta-1} L_{it}^\eta w_t(1 - \alpha)\eta \left(\frac{Y_{it}}{L_{it}} \right)^{-1} \\ p_{bt} &= \left[\frac{w_t}{p_{bt}}(1 - \alpha)\eta \right]^{\eta-1} \left(\frac{L_{pt}}{N_t} \right)^\eta w_t(1 - \alpha)\eta N_t^{\frac{1}{\sigma-1}} \frac{L_{pt}}{Y_t} \\ p_{bt}^\eta &= [w_t(1 - \alpha)\eta]^\eta L_{pt}^{\eta+1} N_t^{\frac{1}{\sigma-1}-\eta} \frac{1}{Y_t} \tag{A.7} \\ p_{bt} &= w_t(1 - \alpha)\eta L_{pt} N_t^{(\frac{1}{\sigma-1}-\eta)\frac{1}{\eta}} \left(\frac{L_{pt}}{Y_t} \right)^{\frac{1}{\eta}} \\ p_{bt} &= (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{(\frac{1}{\sigma-1}-\eta)\frac{1}{\eta}} L_{pt}^{\frac{1}{\eta}} Y_t^{1-\frac{1}{\eta}} \end{aligned}$$

Now, we need to solve Y_t . For that, from above $\frac{Y_{it}}{L_{it}} = N_t^{-\frac{1}{\sigma-1}} \frac{Y_t}{L_{pt}}$ and $L_{it} = \frac{L_{pt}}{N_t}$. Thus:

$$Y_t = N_t^{\frac{\sigma}{\sigma-1}} Y_{it} \tag{A.8}$$

But from the firm's problem we have

$$Y_{it} = D_{it}^\eta L_{it} \tag{A.9}$$

Therefore, solve for D_{it} , where:

$$D_{it} = \alpha Y_{it} + (1 - \alpha) D_{bit} \tag{A.10}$$

But from the data intermediary problem $D_{bit} = B_t$ and from A.5, $D_{bit} = B_t = N_t D_{sit}$,

and by definition $D_{sit} = \tilde{x}_{it}Y_{it}$. Thus:

$$D_{bit} = N_t \tilde{x}_{it} Y_{it} \quad (\text{A.11})$$

Substitute A.11 in A.10 and we get:

$$\begin{aligned} D_{it} &= \alpha Y_{it} + (1 - \alpha) D_{bit} \\ &= \alpha Y_{it} + (1 - \alpha) N_t \tilde{x}_{it} Y_{it} \\ &= Y_{it} (\alpha + (1 - \alpha) N_t \tilde{x}_{it}) \end{aligned} \quad (\text{A.12})$$

Substitute in A.9:

$$\begin{aligned} Y_{it} &= (Y_{it} (\alpha + (1 - \alpha) N_t \tilde{x}_{it}))^\eta L_{it} \\ &= Y_{it}^\eta (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^\eta L_{it} \\ &= Y_{it}^\eta (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^\eta \left(\frac{L_{pt}}{N_t} \right) \\ \Rightarrow Y_{it} &= (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t} \right)^{\frac{1}{1-\eta}} \end{aligned} \quad (\text{A.13})$$

Substitute this expression for Y_{it} in A.8:

$$\begin{aligned} Y_t &= N_t^{\frac{\sigma}{\sigma-1}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t} \right)^{\frac{1}{1-\eta}} \\ &= N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \end{aligned} \quad (\text{A.14})$$

Use this in A.7:

$$\begin{aligned} p_{bt} &= (1 - \alpha) \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\left(\frac{1}{\sigma-1} - \eta \right) \frac{1}{\eta}} L_{pt}^{\frac{1}{\eta}} Y_t^{1 - \frac{1}{\eta}} \\ &= (1 - \alpha) \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\left(\frac{1}{\sigma-1} - \eta \right) \frac{1}{\eta}} L_{pt}^{\frac{1}{\eta}} \left(N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \right)^{1 - \frac{1}{\eta}} \\ &= (1 - \alpha) \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\left(\frac{1}{\sigma-1} - \eta \right) \frac{1}{\eta} + \left(\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta} \right) \left(1 - \frac{1}{\eta} \right)} L_{pt}^{\frac{1}{\eta} + \frac{1}{1-\eta} \left(1 - \frac{1}{\eta} \right)} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\left(\frac{\eta}{1-\eta} \right) \left(1 - \frac{1}{\eta} \right)} \\ &= (1 - \alpha) \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\frac{1}{\sigma-1}} L_{pt}^0 (\alpha + (1 - \alpha) \tilde{x}_{it} N_t)^{-1} \end{aligned} \quad (\text{A.15})$$

Finally,

$$p_{bt} = (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\frac{1}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} \quad (\text{A.16})$$

We will use this in the FOC with respect to L_{it} . Recall from A.3:

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) Y_{it}^{-\frac{1}{\epsilon}} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} B_t^{\frac{1}{\epsilon}} \tilde{x}_{it}^{1-\frac{1}{\epsilon}} + \alpha \left(\frac{p_{bt}}{1 - \alpha}\right) \right] \left(\frac{Y_{it}}{L_{it}}\right) = w_t$$

But from A.6, $p_{bt} = \frac{p_{sit}}{N_t}$ and by the cost minimization problem of the data intermediary, $p_{bt} = \frac{p_{sit}}{N_t} = \frac{1}{N_t} \lambda_{DI} N_t^{-\frac{1}{\epsilon}} \left(\frac{B_t}{\tilde{x}_{it} Y_{it}}\right)^{\frac{1}{\epsilon}}$. Substituting above:

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) p_{bt} \tilde{x}_{it} N_t + \alpha \left(\frac{p_{bt}}{1 - \alpha}\right) \right] \left(\frac{Y_{it}}{L_{it}}\right) = w_t$$

but from above $\frac{Y_{it}}{L_{it}} = N_t^{-\frac{1}{\sigma-1}} \frac{Y_t}{L_{pt}}$ and therefore:

$$\left[\left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + \left(1 - \frac{1}{\epsilon}\right) p_{bt} \tilde{x}_{it} N_t + \alpha \left(\frac{p_{bt}}{1 - \alpha}\right) \right] N_t^{-\frac{1}{\sigma-1}} \frac{Y_t}{L_{pt}} = w_t$$

Also, using A.8:

$$\Rightarrow \left(1 - \frac{1}{\sigma}\right) N_t^{\frac{1}{\sigma-1}} + p_{bt} \left[\left(1 - \frac{1}{\epsilon}\right) \tilde{x}_{it} N_t + \frac{\alpha}{1 - \alpha} \right] = N_t^{\frac{1}{\sigma-1}} \left(\frac{w_t L_{pt}}{Y_t}\right)$$

$$\Rightarrow \left(1 - \frac{1}{\sigma}\right) N_t^{\frac{1}{\sigma-1}} + (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) N_t^{\frac{1}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} \left[\left(1 - \frac{1}{\epsilon}\right) \tilde{x}_{it} N_t + \frac{\alpha}{1 - \alpha} \right] = N_t^{\frac{1}{\sigma-1}} \left(\frac{w_t L_{pt}}{Y_t}\right)$$

$$\Rightarrow \left(1 - \frac{1}{\sigma}\right) + (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} \left[\left(1 - \frac{1}{\epsilon}\right) \tilde{x}_{it} N_t + \frac{\alpha}{1 - \alpha} \right] = \left(\frac{w_t L_{pt}}{Y_t}\right)$$

$$\Rightarrow \left(1 - \frac{1}{\sigma}\right) + (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) \frac{\left(1 - \frac{1}{\epsilon}\right) \tilde{x}_{it} N_t + \frac{\alpha}{1 - \alpha}}{\alpha + (1 - \alpha)\tilde{x}_{it}N_t} = \left(\frac{w_t L_{pt}}{Y_t}\right)$$

Next, define $f(N_t) = \frac{\left(1 - \frac{1}{\epsilon}\right) \tilde{x}_{it} N_t + \frac{\alpha}{1 - \alpha}}{\alpha + (1 - \alpha)\tilde{x}_{it}N_t}$ Thus:

$$\begin{aligned} \Rightarrow \left(1 - \frac{1}{\sigma}\right) + (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) f(N_t) &= \left(\frac{w_t L_{pt}}{Y_t}\right) \\ \Rightarrow \frac{w_t L_{pt}}{Y_t} &= \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} \end{aligned} \quad (\text{A.17})$$

When N_t is large, $\lim_{N_t \rightarrow \infty} f(N_t) = \frac{(1 - \frac{1}{\epsilon})}{(1 - \alpha)}$. Therefore:

$$\Rightarrow \left(\frac{w_t L_{pt}}{Y_t}\right) = \frac{(\sigma - 1)}{\sigma(1 - \eta \frac{\epsilon}{\epsilon - 1})} \quad (\text{A.18})$$

Next, substitute A.14 in A.17 :

$$\begin{aligned} w_t &= \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} \frac{Y_t}{L_{pt}} \\ &= \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha)N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} \end{aligned} \quad (\text{A.19})$$

So far we have:

$$\left\{ \begin{aligned} \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha)N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} &= w_t \\ N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha)N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} &= Y_t \\ (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) N_t^{(\frac{1}{\sigma-1} - \eta)\frac{1}{\eta}} L_{pt}^{\frac{1}{\eta}} Y_t^{1 - \frac{1}{\eta}} &= p_{bt} \\ p_{bt} &= \frac{p_{sit}}{N_t} \\ 1 &= x_{it} \\ Y_{it}(\alpha + (1 - \alpha)N_t \tilde{x}_{it}) &= D_{it} \\ N_t \tilde{x}_{it} Y_{it} &= D_{bit} \\ Y_t N_t^{-\frac{\sigma}{\sigma-1}} &= Y_{it} \\ N_t D_{sit} &= B_t \\ \tilde{x}_{it} Y_{it} &= D_{sit} \\ L_{it} &= \frac{L_{pt}}{N_t} \end{aligned} \right. \quad (\text{A.20})$$

B.5.2 The Value of the Firm and Profits

The firm's problem is:

$$r_t V_{it} = \max_{\{L_{it}, D_{bit}, x_{it}, \tilde{x}_{it}\}} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} - p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it} - \delta(\tilde{x}_{it}) V_{it} + \dot{V}_{it}$$

Thus:

$$V_{it} = \frac{\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} - p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}}$$

Next, define $\pi_{it} = \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w L_{it}$. Then using the expressions above:

$$\begin{aligned} \pi_{it} &= \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w L_{it} \\ &= \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w \left(\frac{L_{pt}}{N_t} \right) \\ &= \left(N_t^{\frac{1}{\sigma-1}} \right) N_t^{-\frac{\sigma}{\sigma-1}} Y_t - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} \left(\frac{L_{pt}}{N_t} \right) \\ &= N_t^{-1} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} \\ &= N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta} - 1} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta} - 1} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} \\ &= (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} \left(1 - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)} \right) \end{aligned} \tag{A.21}$$

On the other hand:

$$\begin{aligned} -p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it} &= - \left(\frac{p_{sit}}{N_t} \right) (N_t \tilde{x}_{it} Y_{it}) + p_{sit} \tilde{x}_{it} Y_{it} \\ &= 0 \end{aligned} \tag{A.22}$$

Hence, the value of the firm is given by:

$$\begin{aligned}
V_{it} &= \frac{\left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} - p_{bt} D_{bit} + p_{sit} \tilde{x}_{it} Y_{it}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \\
&= \frac{(\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(1 - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)}\right) \\
&= \frac{(\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(1 - \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))}\right) \quad (\text{A.23}) \\
&= \frac{(\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(\frac{\sigma(1 - (1 - \alpha)\eta f(N_t)) - (\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))}\right) \\
&= \frac{(\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(\frac{1 - \sigma(1 - \alpha)\eta f(N_t)}{\sigma(1 - (1 - \alpha)\eta f(N_t))}\right)
\end{aligned}$$

B.5.3 The Free Entry Condition

$$\chi w_t = V_{it} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right)$$

Using A.19 and A.23:

$$\begin{aligned}
&\chi \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta} - 1} = \dots \\
&\dots \frac{(\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(1 - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)}\right) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right) \\
\Rightarrow &\chi \frac{(1 - \frac{1}{\sigma}) N_t}{1 - (1 - \alpha)\eta f(N_t)} L_{pt}^{-1} = \frac{1}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(1 - \frac{(1 - \frac{1}{\sigma})}{1 - (1 - \alpha)\eta f(N_t)}\right) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right) \\
\left(\frac{N_t}{L_{pt}}\right) &= \frac{(1 - (1 - \alpha)\eta f(N_t)) \sigma}{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}\right) (\sigma - 1)} \left(1 - \frac{\sigma - 1}{\sigma(1 - (1 - \alpha)\eta f(N_t))}\right) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right)
\end{aligned}$$

$$\begin{aligned}
\Rightarrow \left(\frac{N_t}{L_{pt}} \right) &= \frac{\sigma(1 - (1 - \alpha)\eta f(N_t)) - \sigma + 1}{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}} \right) (\sigma - 1)} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L} \right) \\
\Rightarrow \left(\frac{N_t}{L_{pt}} \right) &= \frac{1 - \sigma(1 - \alpha)\eta f(N_t)}{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}} \right) (\sigma - 1)} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L} \right) \\
\Rightarrow \left(\frac{L_{pt}}{N_t} \right) &= \frac{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}} \right) (\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right) \tag{A.24}
\end{aligned}$$

From the evolution of the number of varieties, we have $\dot{N}_t = \frac{1}{\chi}(L_t - L_{pt})$. Thus:

$$\frac{\dot{N}_t}{N_t} = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$$

In BGP: $\frac{\dot{N}_t}{N_t} = g_N$ and since $g_N = g_L$, then:

$$\begin{aligned}
\Rightarrow g_L &= \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right) \\
\Rightarrow \frac{L_t}{N_t} &= \frac{L_{pt}}{N_t} + \chi g_L
\end{aligned}$$

Next, using $\frac{\dot{V}_{it}}{V_{it}} = g_\pi$ and substituting A.24:

$$\Rightarrow \frac{L_t}{N_t} = \frac{\chi(r + \delta(\tilde{x}_{it}) - g_\pi)(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right) + \chi g_L$$

$$\Rightarrow N_t = \frac{L_t}{\frac{\chi(r + \delta(\tilde{x}_{it}) - g_\pi)(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right) + \chi g_L}$$

Define $\nu(N_t) = \frac{\chi(r + \delta(\tilde{x}_{it}) - g_\pi)(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right)$, so that:

$$N_t = \frac{L_t}{\nu(N_t) + \chi g_L}$$

Moreover, define $\psi(N_t) = \frac{1}{\nu(N_t) + \chi g_L}$ so that:

$$N_t = L_t \psi(N_t) \tag{A.25}$$

B.6 Solution of the Competitive Equilibrium

B.6.1 Data shared with other firms

From above, the FOC for \tilde{x}_{it} is:

$$p_{sit}Y_{it} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it})V_{it} + \mu_{\tilde{x}0} - \mu_{\tilde{x}1} = 0$$

Assume an interior solution so $\mu_{\tilde{x}0} = \mu_{\tilde{x}1} = 0$.

$$p_{sit}Y_{it} \left(1 - \frac{1}{\epsilon}\right) - \delta'(\tilde{x}_{it})V_{it} = 0$$

but from above $p_{bt} = \frac{p_{sit}}{N_t}$ and $p_{bt} = (1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) N_t^{\frac{1}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1}$ Therefore:

$$(1 - \alpha)\eta \left(\frac{w_t L_{pt}}{Y_t}\right) N_t^{\frac{\sigma}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} Y_{it} \frac{(\epsilon - 1)}{\epsilon} - \delta'(\tilde{x}_{it})V_{it} = 0$$

Substitute $\frac{w_t L_{pt}}{Y_t}$ using A.17:

$$(1 - \alpha)\eta \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} N_t^{\frac{\sigma}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} Y_{it} \frac{(\epsilon - 1)}{\epsilon} - \delta'(\tilde{x}_{it})V_{it} = 0$$

Substitute Y_{it} using A.13:

$$(1 - \alpha)\eta \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} N_t^{\frac{\sigma}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} (\alpha + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} \frac{(\epsilon - 1)}{\epsilon} = \delta'(\tilde{x}_{it})V_{it}$$

Next, substitute V_{it} using

$$(1 - \alpha)\eta \frac{(\sigma - 1)}{\sigma(1 - (1 - \alpha)\eta f(N_t))} N_t^{\frac{\sigma}{\sigma-1}} (\alpha + (1 - \alpha)\tilde{x}_{it}N_t)^{-1} (\alpha + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} \frac{(\epsilon - 1)}{\epsilon} = \delta'(\tilde{x}_{it}) \frac{(\alpha + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}}}{r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}}} \left(\frac{1 - \sigma(1 - \alpha)\eta f(N_t)}{\sigma(1 - (1 - \alpha)\eta f(N_t))}\right)$$

Simplify:

$$\begin{aligned}
(1 - \alpha)\eta(\sigma - 1)N_t(\alpha + (1 - \alpha)N_t\tilde{x}_{it})^{-1}\frac{(\epsilon - 1)}{\epsilon} &= \delta'(\tilde{x}_{it})\frac{1 - \sigma(1 - \alpha)\eta f(N_t)}{r + \delta(\tilde{x}_{it}) - g_\pi} \\
\Rightarrow \frac{\delta'(\tilde{x}_{it})}{r + \delta(\tilde{x}_{it}) - g_\pi}(\alpha + (1 - \alpha)N_t\tilde{x}_{it}) &= \frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta N_t\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \\
\Rightarrow \frac{\delta'(\tilde{x}_{it})}{r + \delta(\tilde{x}_{it}) - g_\pi}\left(\frac{\alpha}{N_t} + (1 - \alpha)\tilde{x}_{it}\right) &= \frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)}
\end{aligned}$$

Next, use $\delta'(\tilde{x}_{it}) = \delta_0\tilde{x}_{it}$ and $\delta(\tilde{x}_{it}) = \frac{\delta_0}{2}\tilde{x}_{it}^2$ and N_t is large :

$$\begin{aligned}
&\Rightarrow \frac{(1 - \alpha)\tilde{x}_{it}^2\delta_0}{r + \frac{\delta_0}{2}\tilde{x}_{it}^2 - g_\pi} = \frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \\
&\Rightarrow (1 - \alpha)\tilde{x}_{it}^2\delta_0 = \left(r + \frac{\delta_0}{2}\tilde{x}_{it}^2 - g_\pi\right)\frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \\
&\Rightarrow \tilde{x}_{it}^2\delta_0 \left[(1 - \alpha) - \frac{(\epsilon - 1)}{2\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \right] = (r - g_\pi)\frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \\
&\Rightarrow \tilde{x}_{it} = \left(\frac{(r - g_\pi)\frac{(\epsilon - 1)}{\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)}}{\delta_0(1 - \alpha) - \delta_0\frac{(\epsilon - 1)}{2\epsilon}(1 - \alpha)\eta\frac{(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)}} \right)^{\frac{1}{2}}
\end{aligned}$$

When N_t is large, $\lim_{N_t \rightarrow \infty} f(N_t) = \frac{(1 - \frac{1}{\epsilon})}{(1 - \alpha)}$.

$$\Rightarrow \tilde{x}_{it} = \left(\frac{(r - g_\pi)2(\epsilon - 1)(1 - \alpha)\eta\frac{(\sigma - 1)}{\epsilon - \sigma\eta(\epsilon - 1)}}{2\delta_0(1 - \alpha) - \delta_0(\epsilon - 1)(1 - \alpha)\eta\frac{(\sigma - 1)}{\epsilon - \sigma\eta(\epsilon - 1)}} \right)^{\frac{1}{2}}$$

$$\begin{aligned}\Rightarrow \tilde{x}_{it} &= \left(\frac{(r - g_\pi) 2 (\epsilon - 1) \eta \frac{(\sigma - 1)}{\epsilon - \sigma \eta (\epsilon - 1)}}{2 \delta_0 - (\epsilon - 1) \delta_0 \eta \frac{(\sigma - 1)}{\epsilon - \sigma \eta (\epsilon - 1)}} \right)^{\frac{1}{2}} \\ \Rightarrow \tilde{x}_{it} &= \left(\frac{2 (r - g_\pi) \left[(\epsilon - 1) \eta \frac{(\sigma - 1)}{\epsilon - \sigma \eta (\epsilon - 1)} \right]}{\delta_0 (2 - \left[(\epsilon - 1) \eta \frac{(\sigma - 1)}{\epsilon - \sigma \eta (\epsilon - 1)} \right])} \right)^{\frac{1}{2}}\end{aligned}$$

Define

$$\Gamma = (\epsilon - 1) \eta \frac{(\sigma - 1)}{\epsilon - \sigma \eta (\epsilon - 1)} = \frac{(\sigma - 1)}{\frac{\epsilon}{(\epsilon - 1) \eta} - \sigma} = \frac{\eta (\sigma - 1)}{\frac{\epsilon}{\epsilon - 1} - \sigma}. \quad (\text{A.26})$$

Then:

$$\Rightarrow \tilde{x}_{it} = \left(\frac{2 (r - g_\pi) \Gamma}{\delta_0 (2 - \Gamma)} \right)^{\frac{1}{2}}$$

Since $\rho = r - g_\pi$ then:

$$\Rightarrow \tilde{x}_{it} = \left(\frac{2 \rho \Gamma}{\delta_0 (2 - \Gamma)} \right)^{\frac{1}{2}}. \quad (\text{A.27})$$

B.6.2 Data used by own firm

From A.1 above:

$$x_f = 1$$

B.6.3 Firm Size

Use A.24:

$$L_{it} = \left(\frac{L_{pt}}{N_t} \right) = \frac{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}} \right) (\sigma - 1)}{1 - \sigma(1 - \alpha) \eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right)$$

When N_t is large:

$$\begin{aligned}
L_{it} &= \frac{\chi \left(r + \delta(\tilde{x}_{it}) - \frac{\dot{V}_{it}}{V_{it}} \right) (\sigma - 1)}{1 - \sigma\eta \left(1 - \frac{1}{\epsilon} \right)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right) \\
&= \frac{\chi (\rho + \delta(\tilde{x}_{it})) (\sigma - 1)}{1 - \sigma\eta \left(1 - \frac{1}{\epsilon} \right)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right) \\
&= \chi g_L \frac{\rho + \delta(\tilde{x}_{it})}{g_L + \delta(\tilde{x}_{it})} \cdot \frac{\sigma - 1}{1 - \sigma\eta \left(\frac{\epsilon - 1}{\epsilon} \right)}
\end{aligned}$$

And define

$$\nu_f = \chi g_L \frac{\rho + \delta(\tilde{x}_{it})}{g_L + \delta(\tilde{x}_{it})} \cdot \frac{\sigma - 1}{1 - \sigma\eta \left(\frac{\epsilon - 1}{\epsilon} \right)}$$

These are equations (72) and (75).

B.6.4 Number of Varieties

From A.25, $N_t = L_t \psi(N_t)$ with $\psi(N_t) = \frac{1}{\nu(N_t) + \chi g_L}$ and $\nu(N_t) = \frac{\chi(r + \delta(\tilde{x}_{it}) - g_\pi)(\sigma - 1)}{1 - \sigma(1 - \alpha)\eta f(N_t)} \left(\frac{g_L}{g_L + \delta(\tilde{x}_{it})} \right)$. Now, from above, we know that when N_t is large $\nu = \chi g_L \frac{\rho + \delta(\tilde{x}_{it})}{g_L + \delta(\tilde{x}_{it})} \cdot \frac{\sigma - 1}{1 - \sigma\eta \left(\frac{\epsilon - 1}{\epsilon} \right)}$. Thus:

$$N_t = \frac{1}{\nu_f + \chi g_L} L_t = \psi_f L_t$$

B.6.5 Aggregate output

From A.14, $Y_t = N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha + (1 - \alpha) N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}$. But $\left(\frac{L_{pt}}{N_t} \right)^f = \nu_f$ Therefore:

$$Y_t^f = (N_t)^{\frac{\sigma}{\sigma-1}} (N_t)^{\frac{\eta}{1-\eta}} \left(\frac{\alpha}{N_t} + (1 - \alpha) \tilde{x}_{it} \right)^{\frac{\eta}{1-\eta}} (\nu_f)^{\frac{1}{1-\eta}}$$

When N_t is large:

$$\begin{aligned}
Y_t^f &= (N_t)^{\frac{\sigma}{\sigma-1}} (N_t)^{\frac{\eta}{1-\eta}} ((1 - \alpha) \tilde{x}_{it})^{\frac{\eta}{1-\eta}} (\nu_f)^{\frac{1}{1-\eta}} \\
&= (\nu_f (1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (N_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}}
\end{aligned}$$

and from equation (79), $N_t = \psi_f L_t$, thus:

$$Y_t^f = (\nu_f (1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}}$$

B.6.6 Consumption per capita

$$c_t^f = \frac{Y_t^f}{L_t} \propto L_t^{\frac{1}{\sigma-1} + \frac{\eta}{1-\eta}}$$

This is equation (83) in the paper.

B.6.7 Consumption per capita growth

Using equation (83):

$$g_c^f = \left(\frac{1}{\sigma - 1} + \frac{\eta}{1 - \eta} \right) g_L$$

This is equation (85) in the paper.

B.6.8 Firm production

Combining equation (80) $Y_t^f = (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}}$ and A.8:

$$\begin{aligned} Y_{it}^f &= (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}} N_t^{-\frac{\sigma}{\sigma-1}} \\ &= (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}} (\psi_f L_t)^{-\frac{\sigma}{\sigma-1}} \\ &= (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\eta}{1-\eta}} \end{aligned}$$

and this is equation 91 in the paper.

B.6.9 Data Production

Combining equation 91 and A.12:

$$\begin{aligned} D_{it}^f &= Y_{it}(\alpha + (1 - \alpha)N_t \tilde{x}_{it}) \\ &= (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\eta}{1-\eta}} N_t \left(\frac{\alpha}{N_t} + (1 - \alpha) \tilde{x}_{it} \right) \\ &= (\nu_f(1 - \alpha)^{\eta} \tilde{x}_{it}^{\eta})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\eta}{1-\eta} + 1} (1 - \alpha) \tilde{x}_{it} \\ &= (\nu_f(1 - \alpha) \tilde{x}_{it} \psi_f L_t)^{\frac{1}{1-\eta}} \end{aligned}$$

and this is Equation 87 in the paper.

B.6.10 Aggregate data production

By definition $D_t = N_t D_{it}$, hence:

$$\begin{aligned} D_t^f &= \psi_f L_t (\nu_f(1 - \alpha) \tilde{x}_{it} \psi_f L_t)^{\frac{1}{1-\eta}} \\ &= (\nu_f(1 - \alpha) \tilde{x}_{it})^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{1}{1-\eta} + 1} \end{aligned}$$

and this is exactly equation 89 in the paper.

B.6.11 Labor share

From A.18:

$$\left(\frac{w_t L_{pt}}{Y_t}\right)^f = \frac{(1 - \frac{1}{\sigma})}{1 - \eta \frac{\epsilon}{\epsilon-1}} = \frac{\sigma}{\sigma \left(1 - \eta \frac{\epsilon}{\epsilon-1}\right)}$$

which is equation (93).

B.6.12 Profit share

From A.21, $\pi_t = (\alpha + (1 - \alpha)N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} \left(1 - \frac{(1-\frac{1}{\sigma})}{1-(1-\alpha)\eta f(N_t)}\right)$ and when N_t is large:

$$\begin{aligned} \pi_t &= (\alpha + (1 - \alpha)N_t \tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} \left(\frac{(1 - (1 - \alpha)\eta f(N_t)) \sigma - (\sigma - 1)}{(1 - (1 - \alpha)\eta f(N_t)) \sigma}\right) \\ &= ((1 - \alpha)\tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta} + \frac{\eta}{1-\eta}} \left(\frac{1 - (1 - \alpha)\eta f(N_t) \sigma}{(1 - (1 - \alpha)\eta f(N_t)) \sigma}\right) \\ &= ((1 - \alpha)\tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta} + \frac{\eta}{1-\eta}} \left(\frac{1 - \eta \sigma \left(\frac{\epsilon-1}{\epsilon}\right)}{(1 - \eta \left(\frac{\epsilon-1}{\epsilon}\right)) \sigma}\right) \end{aligned}$$

Hence:

$$\begin{aligned}
\left(\frac{\pi_t N_t}{Y_t}\right)^f &= \frac{((1-\alpha)\tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{1}{\sigma-1} - \frac{1}{1-\eta} + 1 + \frac{\eta}{1-\eta}} \left(\frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma}\right)}{(\nu_f(1-\alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}}} \\
&= \frac{L_{pt}^{\frac{1}{1-\eta}} (\psi_f L_t)^{\frac{1}{1-\eta}} \left(\frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma}\right)}{(\nu_f)^{\frac{1}{1-\eta}}} \\
&= \frac{(\nu_f N_t)^{\frac{1}{1-\eta}} (\psi_f L_t)^{-\frac{1}{1-\eta}} \left(\frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma}\right)}{(\nu_f)^{\frac{1}{1-\eta}}} \\
&= (N_t)^{\frac{1}{1-\eta}} (\psi_f L_t)^{-\frac{1}{1-\eta}} \left(\frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma}\right) \\
&= (\psi_f L_t)^{\frac{1}{1-\eta}} (\psi_f L_t)^{-\frac{1}{1-\eta}} \frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma} \\
&= \frac{1-\eta\sigma(\frac{\epsilon-1}{\epsilon})}{(1-\eta(\frac{\epsilon-1}{\epsilon}))\sigma}
\end{aligned}$$

which is equation (94).

B.6.13 Data Share

Use $p_{at} = \frac{\alpha}{1-\alpha} p_{bt}$ to value the data the firm owns, using the perfect substitutes argument. Then,

$$\begin{aligned}
p_{at} Y_{it} + p_{bt} D_{bt} &= p_{bt} \left(\frac{\alpha}{1-\alpha} Y_{it} + D_{bt} \right) \\
&= \frac{p_{bt}}{1-\alpha} (\alpha Y_{it} + (1-\alpha) D_{bt}) \\
&= \frac{p_{bt}}{1-\alpha} D_{it} \\
&= \frac{p_{bt}}{1-\alpha} [\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t] Y_{it} \\
&= \frac{p_{bt}}{1-\alpha} [\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t] Y_t N_t^{\frac{-\sigma}{\sigma-1}}
\end{aligned}$$

Now recall from (A.14) that $Y_t = N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t]^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}$. Substituting this into the last equation above gives

$$\begin{aligned} p_{at}Y_{it} + p_{bt}D_{bt} &= \frac{p_{bt}}{1-\alpha} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t] N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t]^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{-\frac{\sigma}{\sigma-1}} \\ &= \frac{p_{bt}}{1-\alpha} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t]^{\frac{1}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{-\frac{1}{1-\eta}} \end{aligned}$$

Now recall from (A.15) that $\frac{p_{bt}}{1-\alpha} = \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\frac{1}{\eta(\sigma-1)} - 1} L_{pt}^{\frac{1}{\eta}} Y_t^{1-\frac{1}{\eta}}$ and substitute this in to get

$$p_{at}Y_{it} + p_{bt}D_{bt} = \eta \left(\frac{w_t L_{pt}}{Y_t} \right) N_t^{\frac{1}{\eta(\sigma-1)} - 1} L_{pt}^{\frac{1}{\eta}} Y_t^{1-\frac{1}{\eta}} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t]^{\frac{1}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} N_t^{-\frac{1}{1-\eta}}$$

Again use equation (A.14) for Y_t to substitute for $Y_t^{-\frac{1}{\eta}}$ and simplify the exponents to get

$$p_{at}Y_{it} + p_{bt}D_{bt} = \eta \left(\frac{w_t L_{pt}}{Y_t} \right) \frac{Y_t}{N_t}$$

which gives the data share of GDP as

$$\frac{N_t(p_{at}Y_{it} + p_{bt}D_{bt})}{Y_t} = \eta \left(\frac{w_t L_{pt}}{Y_t} \right)$$

Finally, using the result for the labor share in equation (93), we have our result:

$$\frac{N_t(p_{at}Y_{it} + p_{bt}D_{bt})}{Y_t} = \frac{\eta}{1 - \eta \frac{\epsilon-1}{\epsilon}} \cdot \frac{\sigma-1}{\sigma}$$

which is equation (101).

B.6.14 Price of a variety

From household problem, $p_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}}$, then:

$$\begin{aligned} p_{it}^f &= \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \\ &= \left(\frac{Y_t}{Y_t N_t^{-\frac{\sigma-1}{\sigma}}}\right)^{\frac{1}{\sigma}} \\ &= N_t^{\frac{1}{\sigma-1}} \end{aligned}$$

and from equation (79), $N_t = \psi_f L_t$, thus:

$$p_{it}^f = (\psi_f L_t)^{\frac{1}{\sigma-1}}$$

which is equation (98).

C Competitive Equilibrium When Consumers Own Data

C.1 Household Problem

The household problem is

$$\begin{aligned} U_0 &= \max_{\{c_{it}, x_{it}, \tilde{x}_{it}\}} \int_0^{\infty} e^{-(\tilde{\rho})t} L_0 u(c_t, x_{it}, \tilde{x}_{it}) dt \\ \text{s.t. } c_t &= \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di \right)^{\frac{\sigma}{\sigma-1}} \\ \dot{a}_t &= (r_t - g_L)a_t + w_t - \int_0^{N_t} p_{it} c_{it} di + \int_0^{N_t} x_{it} p_{st}^a c_{it} di + \int_0^{N_t} \tilde{x}_{it} p_{st}^b c_{it} di \\ &= (r_t - g_L)a_t + w_t - \int_0^{N_t} q_{it} c_{it} di \end{aligned}$$

The Hamiltonian of the problem is

$$H(c_{it}, x_{it}, \tilde{x}_{it}, a_t, \mu_t) = u \left(\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di \right)^{\frac{\sigma}{\sigma-1}}, x_{it}, \tilde{x}_{it} \right) + \mu_t \left[(r_t - g_L)a_t + w_t - \int_0^{N_t} q_{it} c_{it} di \right]$$

The FOCs are:

$$\begin{cases} \frac{\partial H}{\partial c_{it}} = 0 \\ \frac{\partial H}{\partial x_{it}} = 0 \\ \frac{\partial H}{\partial \tilde{x}_{it}} = 0 \\ \frac{\partial H}{\partial a_t} = \tilde{\rho}\mu_t - \dot{\mu}_t \end{cases}$$

First, start with $\frac{\partial H}{\partial c_{it}} = 0$:

$$\frac{1}{\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{\sigma}{\sigma-1}}} \frac{\sigma}{\sigma-1} \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} c_{it}^{\frac{-1}{\sigma}} - \mu_t q_{it} = 0$$

$$\Rightarrow c_t^{\frac{1-\sigma}{\sigma}} c_{it}^{\frac{-1}{\sigma}} - \mu_t q_{it} = 0$$

$$\Rightarrow c_t^{\sigma-1} c_{it} = (\mu_t q_{it})^{-\sigma}$$

$$\Rightarrow (c_t^{\sigma-1} c_{it})^{\frac{\sigma-1}{\sigma}} = (\mu_t q_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_{it}^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} (q_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} \int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di = (\mu_t)^{1-\sigma} \int_0^{N_t} q_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_t^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} q_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{\sigma-1}{\sigma} + \frac{(1-\sigma)^2}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} q_{it}^{1-\sigma} di$$

$$\Rightarrow \mu_t = \frac{1}{c_t \left(\int_0^{N_t} q_{it}^{1-\sigma} di\right)^{\frac{1}{1-\sigma}}}$$

Next, define $P_t = \left(\int_0^{N_t} q_{it}^{1-\sigma} di \right)^{\frac{1}{1-\sigma}}$. Thus:

$$\Rightarrow \mu_t = \frac{1}{c_t P_t} \quad (\text{A.28})$$

Next, plug the expression for μ_t in the FOC:

$$\begin{aligned} c_t^{\frac{1-\sigma}{\sigma}} c_{it}^{\frac{-1}{\sigma}} - \frac{1}{c_t P_t} q_{it} &= 0 \\ \Rightarrow c_{it} &= c_t \left(\frac{q_{it}}{P_t} \right)^{-\sigma} \end{aligned}$$

Using the normalization $P_t = 1$ yields :

$$c_{it} = c_t (q_{it})^{-\sigma} \quad (\text{A.29})$$

and therefore we get equation (52) $q_{it} = \left(\frac{c_t}{c_{it}} \right)^{\frac{1}{\sigma}}$.

Next, compute the FOC $\frac{\partial H}{\partial x_{it}} = 0$:

$$\begin{aligned} \frac{\partial H}{\partial x_{it}} &= 0 \\ \frac{\kappa}{N_t} x_{it} &= \mu_t P_{st}^a c_{it} N_t \\ \frac{\kappa}{N_t^2} x_{it} &= \mu_t P_{st}^a c_{it} \end{aligned}$$

Next, compute the FOC $\frac{\partial H}{\partial \tilde{x}_{it}} = 0$:

$$\begin{aligned} \frac{\partial H}{\partial \tilde{x}_{it}} &= 0 \\ \frac{\tilde{\kappa}}{N_t} \tilde{x}_{it} &= \mu_t P_{st}^b c_{it} \end{aligned}$$

Finally, compute the FOC $\frac{\partial H}{\partial a_t} = \tilde{\rho} \mu_t - \dot{\mu}_t$:

$$\begin{aligned} \mu_t (r_t - g_L) &= \tilde{\rho} \mu_t - \dot{\mu}_t \\ \Rightarrow (r_t - g_L) &= \tilde{\rho} - \frac{\dot{\mu}_t}{\mu_t} \end{aligned}$$

But $\mu_t = c_t^{-1}$ then $\frac{\dot{\mu}_t}{\mu_t} = -g_c$. Thus:

$$\Rightarrow (r_t - g_L) = \tilde{\rho} + g_c$$

Next, use A.28 and A.29 in the FOC of x_{it} :

$$\frac{\kappa}{N_t^2} x_{it} = p_{st}^a (q_{it})^{-\sigma}$$

Thus:

$$x_{it} = \frac{N_t^2 p_{st}^a}{\kappa} (q_{it})^{-\sigma} \quad (\text{A.30})$$

Similarly, use A.28 and A.29 in the FOC of \tilde{x}_{it} :

$$\frac{\tilde{\kappa}}{N_t} \tilde{x}_{it} = p_{st}^b$$

$$\tilde{x}_{it} = \frac{N_t}{\tilde{\kappa}} \frac{1}{c_t} p_{st}^b (q_{it})^{-\sigma}$$

$$\Rightarrow \tilde{x}_{it} = \frac{N_t}{\tilde{\kappa}} p_{st}^b (q_{it})^{-\sigma} \quad (\text{A.31})$$

Now, we know that $q_{it} = \left(\frac{c_t}{c_{it}}\right)^{\frac{1}{\sigma}} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}}$ and that $q_{it} = p_{it} - x_{it} p_{st}^a - \tilde{x}_{it} p_{st}^b$ then :

$$p_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \quad (\text{A.32})$$

which is equation (53).

C.2 Firm Problem

The firm problem is given by:

$$r_t V_{it} = \max_{\{L_{it}, D_{ait}, D_{bit}\}} \left[\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] Y_{it} - w_t L_{it} - p_{bt} D_{bit} - p_{at} D_{ait} - \delta(\tilde{x}_{it}) V_{it} + \dot{V}_{it}$$

$$s.t. Y_{it} = D_{it}^\eta L_{it}$$

$$D_{it} = \alpha D_{ait} + (1 - \alpha) D_{bit}$$

$$D_{ait} \geq 0$$

$$D_{bit} \geq 0$$

Start with the FOC w/r to L_{it} :

$$\frac{\partial \mathbb{L}}{\partial L_{it}} = 0$$

$$\Leftrightarrow \left(-\frac{1}{\sigma} \right) Y_t^{\frac{1}{\sigma}} Y_{it}^{\frac{-1}{\sigma} - 1} \frac{\partial Y_{it}}{\partial L_{it}} Y_{it} + \frac{\partial Y_{it}}{\partial L_{it}} \left(\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right) - w_t = 0$$

$$\frac{\partial Y_{it}}{\partial L_{it}} \left[\left(-\frac{1}{\sigma} \right) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = w_t$$

But $\frac{\partial Y_{it}}{\partial L_{it}} = D_{it}^\eta = \frac{Y_{it}}{L_{it}}$ and therefore the FOC is:

$$\frac{Y_{it}}{L_{it}} \left[\left(1 - \frac{1}{\sigma} \right) \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = w_t$$

$$\Rightarrow \frac{Y_{it}}{L_{it}} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = w_t \quad (\text{A.33})$$

Next, the FOC w/r to D_{ait} :

$$\frac{\partial \mathbb{L}}{\partial D_{ait}} = 0$$

$$\Leftrightarrow \left(-\frac{1}{\sigma} \right) Y_t^{\frac{1}{\sigma}} Y_{it}^{\frac{-1}{\sigma} - 1} \frac{\partial Y_{it}}{\partial D_{ait}} Y_{it} + \frac{\partial Y_{it}}{\partial D_{ait}} \left(\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right) - p_{at} = 0$$

$$\frac{\partial Y_{it}}{\partial D_{ait}} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = p_{at}$$

Next, compute the derivative $\frac{\partial Y_{it}}{\partial D_{ait}}$:

$$\begin{aligned}\frac{\partial Y_{it}}{\partial D_{ait}} &= \frac{\partial Y_{it}}{\partial D_{it}} \frac{\partial D_{it}}{\partial D_{ait}} \\ \Rightarrow \frac{\partial Y_{it}}{\partial D_{ait}} &= \eta \frac{Y_{it}}{D_{it}} \alpha\end{aligned}$$

Thus the FOC w/r to D_{ait} is:

$$\eta \alpha \frac{Y_{it}}{D_{it}} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = p_{at} \quad (\text{A.34})$$

Similarly, the FOC w/r to D_{bit} is:

$$\eta(1 - \alpha) \frac{Y_{it}}{D_{it}} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = p_{bt} \quad (\text{A.35})$$

We have computed the 3 FOCs. Now I solve the problem. First, divide A.34 by A.35 to get:

$$p_{at} = \frac{\alpha}{(1 - \alpha)} p_{bt} \quad (\text{A.36})$$

Second, divide A.33 by A.35:

$$\frac{Y_{it}}{L_{it}} \left[\eta(1 - \alpha) \frac{Y_{it}}{D_{it}} \right]^{-1} = \frac{w_t}{p_{bt}}$$

$$\frac{D_{it}}{\eta(1 - \alpha)L_{it}} = \frac{w_t}{p_{bt}}$$

$$\Rightarrow D_{it} = \frac{w_t}{p_{bt}} \eta(1 - \alpha)L_{it}$$

but we know that $Y_{it} = D_{it}^\eta L_{it}$, therefore $L_{it} = \frac{Y_{it}}{D_{it}^\eta}$. Substituting L_{it} :

$$D_{it}^{\eta+1} = \frac{w_t}{p_{bt}} \eta(1 - \alpha) Y_{it}$$

$$D_{it} = \left[\frac{w_t}{p_{bt}} \eta(1 - \alpha) Y_{it} \right]^{\frac{1}{\eta+1}}$$

but $q_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}}$ therefore $Y_{it} = Y_t q_{it}^{-\sigma}$. Substituting:

$$D_{it} = \left[\frac{w_t}{p_{bt}} \eta (1 - \alpha) Y_t q_{it}^{-\sigma} \right]^{\frac{1}{\eta+1}} \quad (\text{A.37})$$

On the other hand, using A.33 :

$$\begin{aligned} D_{it}^{\eta} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] &= w_t \\ \Rightarrow D_{it}^{\eta} \left[\frac{\sigma - 1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] &= w_t \\ \Rightarrow \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} &= \frac{\sigma}{\sigma - 1} \left(\frac{w_t}{D_{it}^{\eta}} - x_{it} p_{st}^a - \tilde{x}_{it} p_{st}^b \right) \\ \Rightarrow Y_{it} &= Y_t \left[\frac{\sigma}{\sigma - 1} \left(\frac{w_t}{D_{it}^{\eta}} - x_{it} p_{st}^a - \tilde{x}_{it} p_{st}^b \right) \right]^{-\sigma} \end{aligned} \quad (\text{A.38})$$

C.3 The 2 Data Intermediary Problems

The 2 data intermediary problems are given by:

$$\begin{aligned} \max_{p_{ait}, D_{cit}^a} \int_0^{N_t} p_{ait} D_{ait} di - \int_0^{N_t} p_{st}^a D_{cit}^a di \\ \text{s.t. } D_{ait} \leq D_{cit}^a \\ p_{ait} \leq p_{ait}^* \end{aligned}$$

and

$$\begin{aligned} \max_{p_{bit}, D_{cit}^b} \int_0^{N_t} p_{bit} D_{bit} di - \int_0^{N_t} p_{st}^b D_{cit}^b di \\ \text{s.t. } D_{bit} \leq B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} D_{cit}^{\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \\ p_{bit} \leq p_{bit}^* \end{aligned}$$

As in the allocation when firms own the data, these problems can be decomposed in a cost minimization problem and a zero profit condition.

C.3.1 The Cost Minimization Problem

The cost minimization problems are:

$$\begin{aligned} \min_{D_{cit}^a} \int_0^{N_t} p_{st}^a D_{cit}^a di \\ s.t. D_{ait} \leq D_{cit}^a \end{aligned}$$

and

$$\begin{aligned} \min_{D_{cit}^b} \int_0^{N_t} p_{st}^b D_{cit}^b di \\ s.t. D_{bit} \leq B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} D_{cit}^{b\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} \end{aligned}$$

The solutions are given by:

$$\begin{cases} D_{ait} = D_{cit}^a \\ D_{bit} = B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} D_{cit}^{b\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} = N_t D_{cit}^b \end{cases} \quad (\text{A.39})$$

where D_{ait} and D_{bit} are given by the firm problem.

C.3.2 The Zero Profit Condition for Each Data Intermediary Problem

By symmetry, the profit for each data intermediary is:

$$\begin{cases} \pi_{ait} = N_t [p_{ait} D_{ait} - p_{st}^a D_{cit}^a] = N_t D_{ait} [p_{ait} - p_{st}^a] \\ \pi_{bit} = N_t [p_{bit} D_{bit} - p_{st}^b D_{cit}^b] = N_t D_{bit} \left[p_{bit} - \frac{p_{st}^b}{N_t} \right] \end{cases}$$

The zero profit condition yields:

$$\begin{cases} p_{ait} = p_{st}^a \\ p_{bit} = \frac{p_{st}^b}{N_t} \end{cases} \quad (\text{A.40})$$

and these are exactly equation (63).

C.4 Market Clearing Condition for Data

Imposing the market clearing condition in the data market, we get:

$$\begin{aligned} D_{cit}^a &= x_{it}c_{it}L_t \\ D_{cit}^b &= \tilde{x}_{it}c_{it}L_t \end{aligned} \tag{A.41}$$

but from the data intermediary problem $D_{bit} = B_t = \left(N_t^{-\frac{1}{\epsilon}} \int_0^{N_t} D_{cit}^{b\frac{\epsilon-1}{\epsilon}} di \right)^{\frac{\epsilon}{\epsilon-1}} = N_t D_{cit}^b$ and $D_{ait} = D_{cit}^a$. Thus the market clearing conditions yield

$$\begin{aligned} D_{ait} &= x_{it}c_{it}L_t \\ D_{bit} &= \tilde{x}_{it}c_{it}L_t N_t \end{aligned} \tag{A.42}$$

and therefore the supply of data is given by:

$$\begin{aligned} D_{it} &= \alpha D_{ait} + (1 - \alpha) D_{bit} \\ &= \alpha x_{it}c_{it}L_t + (1 - \alpha) \tilde{x}_{it}c_{it}L_t N_t \\ &= c_{it}L_t [\alpha x_{it} + (1 - \alpha) \tilde{x}_{it} N_t] \\ &= Y_{it} [\alpha x_{it} + (1 - \alpha) \tilde{x}_{it} N_t] \end{aligned} \tag{A.43}$$

Moreover, the zero profit condition means:

$$\pi_{ait} + \pi_{bit} = 0$$

$$\Rightarrow p_{at} D_{ait} + p_{bt} D_{bit} = p_{st}^a D_{cit}^a + D_{cit}^b p_{st}^b$$

Next, use [A.39](#):

$$\Rightarrow p_{at} D_{cit}^a + p_{bt} N_t D_{cit}^b = p_{st}^a D_{cit}^a + D_{cit}^b p_{st}^b$$

Substitute using [A.42](#):

$$\Rightarrow p_{at} x_{it} c_{it} L_t + p_{bt} N_t \tilde{x}_{it} c_{it} L_t = p_{st}^a x_{it} c_{it} L_t + \tilde{x}_{it} c_{it} L_t p_{st}^b$$

$$\Rightarrow p_{at}x_{it} + p_{bt}N_t\tilde{x}_{it} = p_{st}^a x_{it} + \tilde{x}_{it}p_{st}^b$$

Use A.36:

$$\Rightarrow \frac{\alpha}{(1-\alpha)}p_{bt}x_{it} + p_{bt}N_t\tilde{x}_{it} = p_{st}^a x_{it} + \tilde{x}_{it}p_{st}^b$$

And we finally get:

$$\Rightarrow \frac{p_{bt}}{(1-\alpha)}[\alpha x_{it} + (1-\alpha)N_t\tilde{x}_{it}] = p_{st}^a x_{it} + \tilde{x}_{it}p_{st}^b \quad (\text{A.44})$$

C.5 Summary of Equations and Solution

C.5.1 Summary

Let's summarize all the equations we have so far, from the FOCs of the consumer and firm problems:

$$\left\{ \begin{array}{l}
N_t^{-\frac{\sigma}{\sigma-1}} = \frac{Y_{it}}{Y_t} = \frac{c_{it}}{c_t} = (q_{it})^{-\sigma} \\
x_{it} = \frac{N_t^2 p_{st}^a}{\kappa} (q_{it})^{-\sigma} \\
\tilde{x}_{it} = \frac{N_t}{\tilde{\kappa}} p_{st}^b (q_{it})^{-\sigma} \\
\frac{Y_{it}}{L_{it}} \left[\frac{\sigma-1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = w_t \\
\eta \alpha \frac{Y_{it}}{D_{it}} \left[\frac{\sigma-1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] = p_{at} \\
p_{at} = \frac{\alpha}{(1-\alpha)} p_{bt} \\
D_{it} = \left[\frac{w_t}{p_{bt}} \eta (1-\alpha) Y_t q_{it}^{-\sigma} \right]^{\frac{1}{\eta+1}} \\
Y_{it} = Y_t \left[\frac{\sigma}{\sigma-1} \left(\frac{w_t}{D_{it}^\eta} - x_{it} p_{st}^a - \tilde{x}_{it} p_{st}^b \right) \right]^{-\sigma} \\
D_{ait} = D_{cit}^a \\
D_{bit} = N_t D_{cit}^b \\
p_{at} = p_{st}^a \\
p_{bt} = \frac{p_{st}^b}{N_t} \\
D_{ait} = x_{it} c_{it} L_t \\
D_{bit} = \tilde{x}_{it} c_{it} L_t N_t \\
\frac{p_{bt}}{(1-\alpha)} [\alpha x_{it} + (1-\alpha) N_t \tilde{x}_{it}] = p_{st}^a x_{it} + \tilde{x}_{it} p_{st}^b \\
D_{it} = Y_{it} [\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t]
\end{array} \right. \tag{A.45}$$

C.5.2 Solution for x_{it} and \tilde{x}_{it}

First, use $(q_{it})^{-\sigma} = N_t^{-\frac{\sigma}{\sigma-1}}$ in A.31 and A.30 to get:

$$x_{it} = \frac{p_{st}^a}{\kappa} N_t^{2-\frac{\sigma}{\sigma-1}} \tag{A.46}$$

$$\tilde{x}_{it} = \frac{p_{st}^b}{\tilde{\kappa}} N_t^{1-\frac{\sigma}{\sigma-1}} \tag{A.47}$$

Next, using $N_t^{-\frac{\sigma}{\sigma-1}} = \frac{Y_{it}}{Y_t}$ and A.43 in A.34:

$$\begin{aligned} \eta\alpha \frac{Y_{it}}{D_{it}} \left[\frac{\sigma-1}{\sigma} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it}p_{st}^a + \tilde{x}_{it}p_{st}^b \right] &= p_{at} \\ \Rightarrow \eta\alpha \frac{\left[\frac{\sigma-1}{\sigma} N_t^{\frac{1}{\sigma-1}} + x_{it}p_{st}^a + \tilde{x}_{it}p_{st}^b \right]}{\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t} &= p_{at} \end{aligned} \quad (\text{A.48})$$

Now, use A.46 and A.47 to substitute p_{st}^a and p_{st}^b :

$$\begin{aligned} \eta\alpha \frac{\left[\frac{\sigma-1}{\sigma} N_t^{\frac{1}{\sigma-1}} + x_{it}^2\kappa N_t^{-2+\frac{\sigma}{\sigma-1}} + \tilde{x}_{it}^2\tilde{\kappa} N_t^{-1+\frac{\sigma}{\sigma-1}} \right]}{\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t} &= p_{at} \\ \Rightarrow \eta\alpha \frac{\left[\frac{\sigma-1}{\sigma} N_t^{\frac{1}{\sigma-1}} + x_{it}^2\kappa N_t^{\frac{2-\sigma}{\sigma-1}} + \tilde{x}_{it}^2\tilde{\kappa} N_t^{\frac{1}{\sigma-1}} \right]}{\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t} &= p_{at} \\ \Rightarrow \eta\alpha \frac{N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2\kappa N_t^{-1} + \tilde{x}_{it}^2\tilde{\kappa} \right]}{\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t} &= p_{at} \\ \Rightarrow \eta\alpha \frac{N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2\kappa N_t^{-1} + \tilde{x}_{it}^2\tilde{\kappa} \right]}{\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t} &= p_{at} \end{aligned}$$

Use $p_{at} = \frac{\alpha}{(1-\alpha)}p_{bt}$ and get:

$$\Rightarrow \eta N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2\kappa N_t^{-1} + \tilde{x}_{it}^2\tilde{\kappa} \right] = \frac{p_{bt}}{1-\alpha} (\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t) \quad (\text{A.49})$$

Now, recall from A.44: $\frac{p_{bt}}{(1-\alpha)} [\alpha x_{it} + (1-\alpha)N_t\tilde{x}_{it}] = p_{st}^a x_{it} + \tilde{x}_{it}p_{st}^b$. With this, substitute the RHS in A.49:

$$\eta N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2\kappa N_t^{-1} + \tilde{x}_{it}^2\tilde{\kappa} \right] = p_{st}^a x_{it} + \tilde{x}_{it}p_{st}^b$$

Again, use A.46 and A.47 to substitute p_{st}^a and p_{st}^b :

$$\eta N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2\kappa N_t^{-1} + \tilde{x}_{it}^2\tilde{\kappa} \right] = x_{it}^2\kappa N_t^{\frac{2-\sigma}{\sigma-1}} + \tilde{x}_{it}^2\tilde{\kappa} N_t^{\frac{1}{\sigma-1}}$$

$$\eta N_t^{\frac{1}{\sigma-1}} \left[\frac{\sigma-1}{\sigma} + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa} \right] = N_t^{\frac{1}{\sigma-1}} [x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa}]$$

$$\eta \frac{\sigma-1}{\sigma} + \eta x_{it}^2 \kappa N_t^{-1} + \eta \tilde{x}_{it}^2 \tilde{\kappa} = x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa}$$

$$\eta \frac{\sigma-1}{\sigma} + (\eta-1) x_{it}^2 \kappa N_t^{-1} + (\eta-1) \tilde{x}_{it}^2 \tilde{\kappa} = 0$$

Finally:

$$\Rightarrow x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa} = \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \quad (\text{A.50})$$

On other hand, divide A.46 by A.47:

$$\frac{x_{it}}{\tilde{x}_{it}} = \frac{p_{st}^a N_t \tilde{\kappa}}{p_{st}^b \kappa}$$

Use the zero profit condition for the data intermediary problems:

$$\Rightarrow \frac{x_{it}}{\tilde{x}_{it}} = \frac{p_{at} N_t \tilde{\kappa}}{p_{bt} N_t \kappa}$$

$$\Rightarrow \frac{x_{it}}{\tilde{x}_{it}} = \frac{p_{at} \tilde{\kappa}}{p_{bt} \kappa}$$

Use A.36:

$$\Rightarrow \frac{x_{it}}{\tilde{x}_{it}} = \frac{\alpha \tilde{\kappa}}{(1-\alpha) \kappa} \quad (\text{A.51})$$

Hence, use A.51 in A.50:

$$\tilde{x}_{it}^2 \kappa N_t^{-1} \left(\frac{\alpha \tilde{\kappa}}{(1-\alpha) \kappa} \right)^2 + \tilde{x}_{it}^2 \tilde{\kappa} = \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma}$$

When N_t is large, $\lim_{N_t \rightarrow \infty} N_t^{-1} \left(\frac{\alpha \tilde{\kappa}}{(1-\alpha) \kappa} \right)^2 = 0$, therefore:

$$\tilde{x}^c = \left[\frac{1}{\tilde{\kappa}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right]^{\frac{1}{2}} \quad (\text{A.52})$$

which is equation (65). Moreover, the expression for x^c is:

$$x^c = \frac{\alpha \tilde{\kappa}}{(1-\alpha)\kappa} \tilde{x}^c$$

which is equation (69).

C.5.3 The prices p_{at} and p_{bt}

From A.49:

$$p_{bt} = (1-\alpha)\eta N_t^{\frac{1}{\sigma-1}} \frac{[\frac{\sigma-1}{\sigma} + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa}]}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t)}$$

In the numerator, when N_t is large:

$$\begin{aligned} p_{bt} &= (1-\alpha)\eta N_t^{\frac{1}{\sigma-1}} \frac{[\frac{\sigma-1}{\sigma} + \tilde{x}_{it}^2 \tilde{\kappa}]}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t)} \\ &= (1-\alpha)\eta N_t^{\frac{1}{\sigma-1}} \frac{[\frac{\sigma-1}{\sigma} + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma}]}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t)} \\ &= (1-\alpha)\eta N_t^{\frac{1}{\sigma-1}} \frac{1}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t)} \\ &= (1-\alpha) N_t^{\frac{1}{\sigma-1}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t)} \end{aligned} \quad (\text{A.53})$$

C.5.4 Wage w_t

Divide A.33 by A.34:

$$w_t = p_{at} \frac{D_{it}}{L_{it} \eta \alpha}$$

Substitute D_{it} and L_{it} :

$$w_t = p_{at} \frac{N_t Y_{it} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it} N_t]}{L_{pt} \eta \alpha}$$

But $N_t^{-\frac{\sigma}{\sigma-1}} = \frac{Y_{it}}{Y_t}$, therefore:

$$w_t = p_{at} \frac{N_t^{1-\frac{\sigma}{\sigma-1}} Y_t [\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t]}{L_{pt} \eta \alpha}$$

On the other hand, $p_{at} = \frac{\alpha}{(1-\alpha)} p_{bt} = \alpha \eta N_t^{\frac{1}{\sigma-1}} \frac{[\frac{\sigma-1}{\sigma} + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa}]}{\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t}$. Substitute:

$$w_t = \alpha \eta N_t^{\frac{1}{\sigma-1}} \frac{[\frac{\sigma-1}{\sigma} + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa}]}{\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t} \frac{N_t^{1-\frac{\sigma}{\sigma-1}} Y_t [\alpha x_{it} + (1-\alpha) \tilde{x}_{it} N_t]}{L_{pt} \eta \alpha}$$

$$w_t = \left[\frac{\sigma-1}{\sigma} + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa} \right] \frac{Y_t}{L_{pt}}$$

When N_t is large:

$$w_t = \left[\frac{\sigma-1}{\sigma} + \tilde{x}_{it}^2 \tilde{\kappa} \right] \frac{Y_t}{L_{pt}}$$

Finally, substitute \tilde{x}_{it} :

$$w_t = \left[\frac{\sigma-1}{\sigma} + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right] \frac{Y_t}{L_{pt}}$$

$$\Rightarrow w_t = \frac{\sigma-1}{\sigma(1-\eta)} \frac{Y_t}{L_{pt}} \tag{A.54}$$

C.5.5 Value of the Firm and Profits

By definition, similarly to the case when firms own the data, the value of the firm is given by:

$$V_t = \frac{\left[\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] Y_{it} - w_t L_{it} - p_{bt} D_{bit} - p_{at} D_{ait}}{r + \delta(\tilde{x}_{it}) - g_V}$$

Let's compute each component separately:

$$V_t = \frac{\overbrace{\left[\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right]}^C Y_{it} - \overbrace{w_t L_{it} - p_{bt} D_{bit} - p_{at} D_{ait}}^{B \quad A}}}{r + \delta(\tilde{x}_{it}) - g_V}$$

First, let's compute A :

$$\begin{aligned}
A &= p_{bt}D_{bit} + p_{at}D_{ait} \\
&= p_{bt} \left[D_{bit} + \frac{\alpha}{1-\alpha} D_{ait} \right] \\
&= \frac{p_{bt}}{(1-\alpha)} [(1-\alpha)D_{bit} + \alpha D_{ait}] \\
&= \frac{p_{bt}}{(1-\alpha)} [\alpha x_{it}c_{it}L_t + (1-\alpha)\tilde{x}_{it}c_{it}L_tN_t] \\
&= \frac{p_{bt}}{(1-\alpha)} Y_{it} [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t] \\
&= \frac{p_{bt}}{(1-\alpha)} N_t^{-\frac{\sigma}{\sigma-1}} Y_t [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t] \\
&= (1-\alpha) N_t^{\frac{1}{\sigma-1}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t)} \frac{1}{(1-\alpha)} N_t^{-\frac{\sigma}{\sigma-1}} Y_t [\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t] \\
&= \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{Y_t}{N_t}
\end{aligned}$$

Note, this term is equation (100), which gives the data share of the economy.

Next, let's compute B :

$$\begin{aligned}
B &= w_t L_{it} \\
&= \frac{\sigma-1}{\sigma(1-\eta)} \frac{Y_t}{L_{pt}} L_{it} \\
&= \frac{\sigma-1}{\sigma(1-\eta)} \frac{Y_t}{L_{pt}} \frac{L_{pt}}{N_t} \\
&= \frac{\sigma-1}{\sigma(1-\eta)} \frac{Y_t}{N_t}
\end{aligned}$$

Finally, compute C :

$$\begin{aligned}
C &= \left[\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] Y_{it} \\
&= \left[N_t^{\frac{1}{\sigma-1}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] Y_{it} \\
&= \left[N_t^{\frac{1}{\sigma-1}} + x_{it}^2 \kappa N_t^{-2+\frac{\sigma}{\sigma-1}} + \tilde{x}_{it}^2 \tilde{\kappa} \right] Y_{it} \\
&= N_t^{\frac{1}{\sigma-1}} \left[1 + x_{it}^2 \kappa N_t^{-1} + \tilde{x}_{it}^2 \tilde{\kappa} \right] Y_{it} \\
&= N_t^{\frac{1}{\sigma-1}} \left[1 + \tilde{x}_{it}^2 \tilde{\kappa} \right] Y_{it} \\
&= N_t^{\frac{1}{\sigma-1}} \left[1 + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right] Y_{it} \\
&= N_t^{\frac{1}{\sigma-1}} \left[1 + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right] Y_t N_t^{-\frac{\sigma}{\sigma-1}} \\
&= \frac{Y_t}{N_t} \left[1 + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right]
\end{aligned}$$

Using (A), (B), and (C) we can compute the value of the firm:

$$\begin{aligned}
V_{it} &= \frac{\overbrace{\left[\left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b \right] Y_{it}}^C - \overbrace{w_t L_{it} - p_{bt} D_{bit} - p_{at} D_{ait}}^{B \quad A}}{r + \delta(\tilde{x}_{it}) - g_V}} \\
&= \frac{1}{r + \delta(\tilde{x}_{it}) - g_V} \left[\frac{Y_t}{N_t} \left[1 + \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \right] - \frac{\sigma-1}{\sigma(1-\eta)} \frac{Y_t}{N_t} - \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{Y_t}{N_t} \right] \\
&= \frac{Y_t}{N_t} \frac{1}{(r + \delta(\tilde{x}_{it}) - g_V)} \frac{(1 - \sigma\eta)}{\sigma(1-\eta)}
\end{aligned}$$

C.6 Solution of the Competitive Equilibrium

C.6.1 Firm Size

Next, impose the free entry condition:

$$\chi w_t = V_{it} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L} \right)$$

Substitute V_{it} and w_t :

$$\begin{aligned}\chi \frac{\sigma - 1}{\sigma(1 - \eta)} \frac{Y_t}{L_{pt}} &= \frac{Y_t}{N_t} \frac{1}{(r + \delta(\tilde{x}_{it}) - g_V)} \frac{(1 - \sigma\eta)}{\sigma(1 - \eta)} \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right) \\ \Rightarrow \frac{L_{pt}}{N_t} &= \chi \frac{(\sigma - 1)(r + \delta(\tilde{x}_{it}) - g_V)}{(1 - \sigma\eta) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right)}\end{aligned}$$

Use $\rho = r - g_V$ on BGP:

$$\begin{aligned}\Rightarrow \frac{L_{pt}}{N_t} &= \chi \frac{(\sigma - 1)(\rho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right)} \\ \Rightarrow \frac{L_{pt}}{N_t} &= \chi \frac{(\sigma - 1)(\rho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta) \left(1 + \frac{\delta(\tilde{x}_{it})}{g_L}\right)} \\ \Rightarrow \frac{L_{pt}}{N_t} &= \chi g_L \frac{(\sigma - 1)(\rho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta)(g_L + \delta(\tilde{x}_{it}))}\end{aligned}\tag{A.55}$$

Define $\nu_c = \frac{L_{pt}}{N_t}$. Finally:

$$\Rightarrow \nu_c = \chi g_L \frac{(\sigma - 1)(\rho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta)(g_L + \delta(\tilde{x}_{it}))}$$

which is equation (74) in the paper.

C.6.2 Number of Varieties

From the evolution of the number of varieties, we have $\dot{N}_t = \frac{1}{\chi}(L_t - L_{pt})$. Thus:

$$\frac{\dot{N}_t}{N_t} = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$$

In BGP: $\frac{\dot{N}_t}{N_t} = g_N$ and since $g_N = g_L$, then:

$$\Rightarrow g_L = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$$

$$\Rightarrow \frac{L_t}{N_t} = \frac{L_{pt}}{N_t} + \chi g_L$$

Next, using $\frac{\dot{V}_{it}}{V_{it}} = g_\pi$ and substituting $\frac{L_{pt}}{N_t}$ using A.55:

$$\Rightarrow \frac{L_t}{N_t} = \chi g_L \frac{(\sigma - 1)(\varrho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta)(g_L + \delta(\tilde{x}_{it}))} + \chi g_L$$

$$\Rightarrow N_t = \frac{L_t}{\chi g_L \frac{(\sigma - 1)(\varrho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta)(g_L + \delta(\tilde{x}_{it}))} + \chi g_L}$$

Define $\nu_c = \chi g_L \frac{(\sigma - 1)(\varrho + \delta(\tilde{x}_{it}))}{(1 - \sigma\eta)(g_L + \delta(\tilde{x}_{it}))}$, so that:

$$N_t = \frac{L_t}{\nu_c + \chi g_L}$$

Moreover, define $\psi_c = \frac{1}{\nu_c + \chi g_L}$ so that:

$$N_t = \psi_c L_t$$

which is equation (79) for the allocation when consumers own the data.

C.6.3 Aggregate Output

First, substitute $D_{it} = Y_{it} [\alpha x_{it} + (1 - \alpha)\tilde{x}_{it}N_t]$ in $Y_{it} = D_{it}^\eta L_{it}$.

$$\begin{aligned} Y_{it} &= (Y_{it}(\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it}))^\eta L_{it} \\ &= Y_{it}^\eta (\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it})^\eta L_{it} \\ &= Y_{it}^\eta (\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it})^\eta \left(\frac{L_{pt}}{N_t}\right) \\ \Rightarrow Y_{it} &= (\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} \end{aligned} \tag{A.56}$$

On the other hand, aggregate output is $Y_t = Y_{it} N_t^{\frac{\sigma}{\sigma-1}}$. Hence:

$$\begin{aligned} Y_t &= N_t^{\frac{\sigma}{\sigma-1}} (\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} \\ &= N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha x_{it} + (1 - \alpha)N_t\tilde{x}_{it})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \end{aligned}$$

When N_t is large:

$$\begin{aligned} Y_t^c &= (N_t)^{\frac{\sigma}{\sigma-1}} (N_t)^{\frac{\eta}{1-\eta}} ((1 - \alpha)\tilde{x}_{it})^{\frac{\eta}{1-\eta}} (\nu_c)^{\frac{1}{1-\eta}} \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (N_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}} \end{aligned}$$

and from equation (79), $N_t = \psi_c L_t$, thus:

$$Y_t^c = (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}}$$

C.6.4 Firm production

$$\begin{aligned} Y_{it}^c &= Y_t N_t^{-\frac{\sigma}{\sigma-1}} \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}} N_t^{-\frac{\sigma}{\sigma-1}} \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\sigma}{\sigma-1} + \frac{\eta}{1-\eta}} (\psi_f L_t)^{-\frac{\sigma}{\sigma-1}} \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\eta}{1-\eta}} \end{aligned}$$

which is equation (91) for allocation c .

C.6.5 Consumption per capita and growth

Consumption per capita is defined as $c_t^c = \frac{Y_t^c}{L_t}$, and using the expression for aggregate output, then:

$$c_t^c = \frac{Y_t^c}{L_t} \propto L_t^{\frac{1}{\sigma-1} + \frac{\eta}{1-\eta}}$$

which is equation (83) in the paper for the allocation c . Using this, the growth is:

$$g_c^c = \left(\frac{1}{\sigma-1} + \frac{\eta}{1-\eta} \right) g_L$$

which is equation (85) in the paper for the allocation c .

C.6.6 Data used by each firm and aggregate data

$$\begin{aligned} D_{it}^c &= Y_{it}(\alpha x_{it} + (1 - \alpha)N_t \tilde{x}_{it}) \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\eta}{1-\eta}} N_t \left(\frac{\alpha}{N_t} x_{it} + (1 - \alpha) \tilde{x}_{it} \right) \\ &= (\nu_c(1 - \alpha)^\eta \tilde{x}_{it}^\eta)^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{\eta}{1-\eta} + 1} (1 - \alpha) \tilde{x}_{it} \\ &= (\nu_c(1 - \alpha) \tilde{x}_{it} \psi_f L_t)^{\frac{1}{1-\eta}} \end{aligned}$$

By definition $D_t = N_t D_{it}$, hence:

$$\begin{aligned} D_t^c &= \psi_c L_t (\nu_c(1 - \alpha) \tilde{x}_{it} \psi_c L_t)^{\frac{1}{1-\eta}} \\ &= (\nu_c(1 - \alpha) \tilde{x}_{it})^{\frac{1}{1-\eta}} (\psi_c L_t)^{\frac{1}{1-\eta} + 1} \end{aligned}$$

and this is equation (89).

C.6.7 Labor Share

From A.54:

$$\left(\frac{w_t L_{pt}}{Y_t}\right)^c = \frac{\sigma - 1}{\sigma(1 - \eta)} \frac{Y_t}{L_{pt}} \frac{L_{pt}}{Y_t} = \frac{\sigma - 1}{\sigma(1 - \eta)}$$

and this is equation (93) in the paper.

C.6.8 Profit share

Using the results from C.5.5, the profit is defined as $\pi_t = C - A - B$, where C , B and A are defined in C.5.5. Then:

$$\begin{aligned} \left(\frac{N_t \pi_t}{Y_t}\right) &= \frac{N_t}{Y_t} \left(\frac{Y_t}{N_t} \left[1 + \frac{\eta}{1 - \eta} \frac{\sigma - 1}{\sigma}\right] - \frac{\sigma - 1}{\sigma(1 - \eta)} \frac{Y_t}{N_t} - \frac{\eta}{1 - \eta} \frac{\sigma - 1}{\sigma} \frac{Y_t}{N_t}\right) \\ &= \frac{(1 - \sigma\eta)}{\sigma(1 - \eta)} \end{aligned}$$

C.6.9 Prices

From A.45, $(q_{it})^{-\sigma} = N_t^{-\frac{\sigma}{\sigma-1}}$. Then:

$$\begin{aligned} q_{it} &= N_t^{\frac{1}{\sigma-1}} \\ &= (L_t \psi_c)^{\frac{1}{\sigma-1}} \end{aligned}$$

Next, from equation (53) in the paper, p_{it} is defined as $p_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} + x_{it} p_{st}^a + \tilde{x}_{it} p_{st}^b$.

Now, from the computation of C in C.5.5, we have $p_{it} = N_t^{\frac{1}{\sigma-1}} \left[1 + \frac{\eta}{1 - \eta} \frac{\sigma - 1}{\sigma}\right]$. Therefore:

$$\begin{aligned} p_{it} &= N_t^{\frac{1}{\sigma-1}} \left[1 + \frac{\eta}{1 - \eta} \frac{\sigma - 1}{\sigma}\right] \\ &= (L_t \psi_c)^{\frac{1}{\sigma-1}} \left[1 + \frac{\eta}{1 - \eta} \frac{\sigma - 1}{\sigma}\right] \end{aligned}$$

C.6.10 Price p_{st}^b

Recall from A.53 that $p_{bt} = (1 - \alpha)N_t^{\frac{1}{\sigma-1}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t)}$ and we know that, by the zero profit condition in the data intermediary, $p_{bt} = \frac{p_{st}^b}{N_t}$. Then:

$$\begin{aligned}
 p_{st}^b &= p_{bt}N_t \\
 &= (1 - \alpha)N_t^{\frac{1}{\sigma-1}+1} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(\alpha x_{it} + (1-\alpha)\tilde{x}_{it}N_t)} \\
 &= (1 - \alpha)N_t^{\frac{1}{\sigma-1}+1} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{N_t \left(\frac{\alpha x_{it}}{N_t} + (1-\alpha)\tilde{x}_{it} \right)} \\
 &= (1 - \alpha)N_t^{\frac{1}{\sigma-1}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{(1-\alpha)\tilde{x}_{it}} \\
 &= N_t^{\frac{1}{\sigma-1}} \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{\tilde{x}_{it}} \\
 &= \frac{\eta}{1-\eta} \frac{\sigma-1}{\sigma} \frac{1}{\tilde{x}_c} (L_t \psi_c)^{\frac{1}{\sigma-1}}
 \end{aligned}$$

D Competitive Equilibrium With Outlaw Data Sharing

D.1 Household Problem

The problem is

$$\begin{aligned}
 U_0 &= \max_{\{c_{it}\}} \int_0^\infty e^{-(\tilde{r})t} L_0 u(c_t, x_{it}) dt \\
 s.t. \quad c_t &= \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di \right)^{\frac{\sigma}{\sigma-1}} \\
 \dot{a}_t &= (r_t - g_L)a_t + w_t - \int_0^{N_t} p_{it} c_{it} di
 \end{aligned}$$

Define the current value Hamiltonian with state variable a_t , control variable c_{it} , and co-state variable μ_t :

$$H(a_t, c_{it}, \mu_t) = u \left(\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di \right)^{\frac{\sigma}{\sigma-1}}, x_{it}, \tilde{x}_{it} \right) + \mu_t \left[(r_t - g_L)a_t + w_t - \int_0^{N_t} p_{it} c_{it} di \right]$$

The FOCs are:

$$\begin{cases} \frac{\partial H}{\partial c_{it}} = 0 \\ \frac{\partial H}{\partial a_t} = \tilde{\rho}\mu_t - \dot{\mu}_t \end{cases}$$

First, start with $\frac{\partial H}{\partial c_{it}} = 0$:

$$\frac{1}{\left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{\sigma}{\sigma-1}}} \frac{\sigma}{\sigma-1} \left(\int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di\right)^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} c_{it}^{\frac{-1}{\sigma}} - \mu_t p_{it} = 0$$

$$\Rightarrow c_t^{\frac{1-\sigma}{\sigma}} c_{it}^{\frac{-1}{\sigma}} - \mu_t p_{it} = 0$$

$$\Rightarrow c_t^{\sigma-1} c_{it} = (\mu_t p_{it})^{-\sigma}$$

$$\Rightarrow (c_t^{\sigma-1} c_{it})^{\frac{\sigma-1}{\sigma}} = (\mu_t p_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_{it}^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} (p_{it})^{1-\sigma}$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} \int_0^{N_t} c_{it}^{\frac{\sigma-1}{\sigma}} di = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{(1-\sigma)^2}{\sigma}} c_t^{\frac{\sigma-1}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow c_t^{\frac{\sigma-1}{\sigma} + \frac{(1-\sigma)^2}{\sigma}} = (\mu_t)^{1-\sigma} \int_0^{N_t} p_{it}^{1-\sigma} di$$

$$\Rightarrow \mu_t = \frac{1}{c_t \left(\int_0^{N_t} p_{it}^{1-\sigma} di\right)^{\frac{1}{1-\sigma}}}$$

Next, use the fact that the price of c_t is normalized to 1, i.e., $P_t = \left(\int_0^{N_t} p_{it}^{1-\sigma} di\right)^{\frac{1}{1-\sigma}} =$

1.

Next, plug the expression for μ_t in the FOC and this gives equation (39):

$$c_{it} = c_t p_{it}^{-\sigma}$$

Next, compute the FOC $\frac{\partial H}{\partial a_t} = \tilde{\rho}\mu_t - \dot{\mu}_t$:

$$\mu_t(r_t - g_L) = \tilde{\rho}\mu_t - \dot{\mu}_t$$

$$\Rightarrow (r_t - g_L) = \tilde{\rho} - \frac{\dot{\mu}_t}{\mu_t}$$

But $\mu_t = c_t^{-1}$ then $\frac{\dot{\mu}_t}{\mu_t} = -g_c$. Thus:

$$\Rightarrow (r_t - g_L) = \tilde{\rho} + g_c$$

D.2 Firm Problem

D.2.1 Define the Problem of the Firm

The firm problem is:

$$r_t V_{it} = \max_{\{L_{it}, D_{bit}\}} \left(\frac{Y_t}{Y_{it}} \right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} + \dot{V}_{it}$$

$$s.t. Y_{it} = D_{it}^\eta L_{it}$$

$$D_{it} = \alpha x_{it} Y_{it}$$

$$x_{it} \in [0; \bar{x}]$$

D.2.2 Compute FOC

To solve this problem, write the Lagrangean:

$$\mathbb{L} = (Y_t)^{\frac{1}{\sigma}} Y_{it}^{1-\frac{1}{\sigma}} - w_t L_{it} + \dot{V}_{it} + \mu_{x0}(x_{it}) + \mu_{x1}(\bar{x} - x_{it})$$

Now take the FOCs. Start with the FOC w/r to L_{it} :

$$\frac{\partial \mathbb{L}}{\partial L_{it}} = 0$$

$$\Leftrightarrow \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \frac{\partial Y_{it}}{\partial L_{it}} = w_t$$

And using $Y_{it} = D_{it}^\eta L_{it}$ and assuming D_{it} depends on L_{it} , then by implicit derivation:

$$\begin{aligned}\frac{\partial Y_{it}}{\partial L_{it}} &= \eta D_{it}^{\eta-1} L_{it} \frac{\partial D_{it}}{\partial L_{it}} + D_{it}^\eta \\ \Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} &= \eta \frac{Y_{it}}{D_{it}} \frac{\partial D_{it}}{\partial L_{it}} + \frac{Y_{it}}{L_{it}}\end{aligned}$$

Next, compute $\frac{\partial D_{it}}{\partial L_{it}}$ using $D_{it} = \alpha x_{it} Y_{it}$:

$$\frac{\partial D_{it}}{\partial L_{it}} = \alpha x_{it} \frac{\partial Y_{it}}{\partial L_{it}}$$

Substituting above:

$$\begin{aligned}\Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} &= \eta \frac{Y_{it}}{D_{it}} \alpha x_{it} \frac{\partial Y_{it}}{\partial L_{it}} + \frac{Y_{it}}{L_{it}} \\ \Rightarrow \frac{\partial Y_{it}}{\partial L_{it}} &= \frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}}\end{aligned}$$

The FOC for L_{it} is then:

$$\frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha x_{it}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} = w_t \quad (\text{A.57})$$

Next, compute the FOC w/r to x_{it} :

$$\begin{aligned}\frac{\partial \mathbb{L}}{\partial x_{it}} &= 0 \\ \Leftrightarrow \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \frac{\partial Y_{it}}{\partial x_{it}} + \mu_{x0} - \mu_{x1} &= 0 \\ \frac{\partial Y_{it}}{\partial x_{it}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} &= -\mu_{x0} + \mu_{x1}\end{aligned}$$

Now to compute $\frac{\partial Y_{it}}{\partial x_{it}}$ use $Y_{it} = D_{it}^\eta L_{it}$:

$$\frac{\partial Y_{it}}{\partial x_{it}} = \eta \frac{Y_{it}}{D_{it}} \frac{\partial D_{it}}{\partial x_{it}}$$

and using $D_{it} = \alpha x_{it} Y_{it}$ and implicit derivation:

$$\frac{\partial D_{it}}{\partial x_{it}} = \alpha x_{it} \frac{\partial Y_{it}}{\partial x_{it}} + Y_{it} \alpha$$

Thus:

$$\begin{aligned} \Rightarrow \frac{\partial Y_{it}}{\partial x_{it}} &= \eta \frac{Y_{it}}{D_{it}} \left[\alpha x_{it} \frac{\partial Y_{it}}{\partial x_{it}} + Y_{it} \alpha \right] \\ \Rightarrow \frac{\partial Y_{it}}{\partial x_{it}} &= \frac{\eta \frac{Y_{it}}{D_{it}} Y_{it} \alpha}{1 - \alpha x_{it} \eta \frac{Y_{it}}{D_{it}}} \end{aligned}$$

Then the FOC w/r to x_{it} is:

$$\frac{\eta \frac{Y_{it}}{D_{it}} Y_{it} \alpha}{1 - \alpha x_{it} \eta \frac{Y_{it}}{D_{it}}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} = -\mu_{x0} + \mu_{x1}$$

Now, note that the LHS is > 0 , then:

$$\mu_{x1} > \mu_{x0} \geq 0$$

$$\Rightarrow \mu_{x1} > 0$$

$$\Rightarrow x_{it} = \bar{x} \tag{A.58}$$

and the FOC w/r to x_{it} is:

$$\frac{\eta \frac{Y_{it}}{D_{it}} Y_{it} \alpha}{1 - \alpha x_{it} \eta \frac{Y_{it}}{D_{it}}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} = \mu_{x1} \tag{A.59}$$

D.3 Free Entry and the Creation of New Varieties

The free entry condition is given by:

$$\chi w_t = V_{it}$$

D.4 Equilibrium with Outlaw Data Sharing

D.4.1 Expressions for Aggregate Output and Firm Output

Now, we need to solve Y_t . For that, from above $\frac{Y_{it}}{L_{it}} = N_t^{-\frac{1}{\sigma-1}} \frac{Y_t}{L_{pt}}$ and $L_{it} = \frac{L_{pt}}{N_t}$. Thus:

$$Y_t = N_t^{\frac{\sigma}{\sigma-1}} Y_{it} \quad (\text{A.60})$$

But from the firm's problem we have

$$Y_{it} = D_{it}^\eta L_{it} \quad (\text{A.61})$$

and $D_{it} = \alpha \bar{x} Y_{it}$. Hence:

$$\begin{aligned} Y_{it} &= (\alpha \bar{x} Y_{it})^\eta L_{it} \\ &= Y_{it}^\eta (\alpha \bar{x})^\eta L_{it} \\ &= Y_{it}^\eta (\alpha \bar{x})^\eta \left(\frac{L_{pt}}{N_t} \right) \\ \Rightarrow Y_{it} &= (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t} \right)^{\frac{1}{1-\eta}} \end{aligned} \quad (\text{A.62})$$

Substitute this expression for Y_{it} in A.60:

$$\begin{aligned} Y_t &= N_t^{\frac{\sigma}{\sigma-1}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t} \right)^{\frac{1}{1-\eta}} \\ &= N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \end{aligned} \quad (\text{A.63})$$

D.4.2 Expression for w_t

From A.57, $\frac{Y_{it}}{L_{it}} \frac{1}{1-\eta} \frac{1}{D_{it} \alpha x_{it}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} = w_t$, hence:

$$\begin{aligned}
w_t &= \frac{\frac{Y_{it}}{L_{it}}}{1 - \eta \frac{Y_{it}}{D_{it}} \alpha \bar{x}} \left(1 - \frac{1}{\sigma}\right) \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \\
&= \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} L_{it}^{-1}}{1 - \eta \frac{Y_{it}}{\alpha \bar{x} Y_{it}} \alpha \bar{x}} \left(1 - \frac{1}{\sigma}\right) N_t^{\frac{1}{\sigma-1}} \\
&= \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} L_{it}^{-1}}{1 - \eta} \left(1 - \frac{1}{\sigma}\right) N_t^{\frac{1}{\sigma-1}} \\
&= \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right)^{\frac{1}{1-\eta}} N_t L_{pt}^{-1}}{1 - \eta} \left(1 - \frac{1}{\sigma}\right) N_t^{\frac{1}{\sigma-1}} \\
&= \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}}}{1 - \eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}}
\end{aligned}$$

Finally:

$$w_t = \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}}}{1 - \eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} \quad (\text{A.64})$$

D.4.3 Expressions for Profits and Value of the Firm

The firm's problem is:

$$r_t V_{it} = \max_{\{L_{it}, D_{bit}, x_{it}, \tilde{x}_{it}\}} \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it} + \dot{V}_{it}$$

Thus:

$$V_{it} = \frac{\left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it}}{r - \frac{\dot{V}_{it}}{V_{it}}}$$

Next, define $\pi_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it}$. Then using the expressions above:

$$\begin{aligned}
\pi_{it} &= \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - wL_{it} \\
&= \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w \left(\frac{L_{pt}}{N_t}\right) \\
&= \left(N_t^{\frac{1}{\sigma-1}}\right) N_t^{-\frac{\sigma}{\sigma-1}} Y_t - \frac{(\alpha\bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} \left(\frac{L_{pt}}{N_t}\right) \\
&= N_t^{-1} Y_t - \frac{(\alpha\bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} \\
&= N_t^{-1} N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha\bar{x})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{\eta}{1-\eta}} - \frac{(\alpha\bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} \\
&= L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} (\alpha\bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]
\end{aligned} \tag{A.65}$$

Then, the value of the firm is given by:

$$\begin{aligned}
V_{it} &= \frac{\left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} Y_{it} - w_t L_{it}}{r - \frac{\dot{V}_{it}}{V_{it}}} \\
&= \frac{L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} (\alpha\bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{r - g_V}
\end{aligned} \tag{A.66}$$

D.5 Solution of the Competitive Equilibrium

D.5.1 Firm Size v_{os}

Use the free entry condition:

$$\chi w_t = V_{it}$$

Substituting the expressions for w_t and V_{it} using [A.64](#) and [A.66](#):

$$\chi \frac{(\alpha\bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} = \frac{L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} (\alpha\bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{r - g_V}$$

$$\chi \frac{1}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{-1} N_t = \frac{\left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{r - g_V}$$

$$\Rightarrow \frac{L_{pt}}{N_t} = \frac{\chi \frac{1}{1-\eta} (1 - \frac{1}{\sigma}) (r - g_V)}{1 - \frac{\sigma-1}{\sigma(1-\eta)}}$$

$$\Rightarrow \frac{L_{pt}}{N_t} = \frac{\chi \frac{1}{1-\eta} (\frac{\sigma-1}{\sigma}) (r - g_V)}{\frac{\sigma(1-\eta) - (\sigma-1)}{\sigma(1-\eta)}}$$

$$\Rightarrow \frac{L_{pt}}{N_t} = \frac{\chi (\sigma - 1) (r - g_V)}{\sigma(1 - \eta) - (\sigma - 1)}$$

but $\rho = r - g_V$ then:

$$\Rightarrow \frac{L_{pt}}{N_t} = \frac{\chi \rho (\sigma - 1)}{1 - \sigma \eta}$$

Define $v_{os} = \frac{L_{pt}}{N_t}$. Then:

$$v_{os} = \frac{\chi \rho (\sigma - 1)}{1 - \sigma \eta} \tag{A.67}$$

which is equation (76).

D.5.2 Number of varieties

From the evolution of the number of varieties, we have $\dot{N}_t = \frac{1}{\chi}(L_t - L_{pt})$. Thus:

$$\frac{\dot{N}_t}{N_t} = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$$

In BGP: $\frac{\dot{N}_t}{N_t} = g_N$ and since $g_N = g_L$, then:

$$\Rightarrow g_L = \frac{1}{\chi} \left(\frac{L_t}{N_t} - \frac{L_{pt}}{N_t} \right)$$

$$\Rightarrow \frac{L_t}{N_t} = \frac{L_{pt}}{N_t} + \chi g_L$$

Next, using $\frac{\dot{V}_{it}}{V_{it}} = g_\pi$ and substituting the firm size:

$$\Rightarrow \frac{L_t}{N_t} = \frac{\chi \rho (\sigma - 1)}{1 - \sigma \eta} + \chi g_L$$

$$\Rightarrow N_t = \frac{L_t}{\frac{\chi\rho(\sigma-1)}{1-\sigma\eta} + \chi g_L}$$

Define $\psi_{os} = \frac{1}{\frac{\chi\rho(\sigma-1)}{1-\sigma\eta} + \chi g_L} = \frac{1}{v_{os} + \chi g_L}$ so that:

$$N_t = L_t \psi_{os} \tag{A.68}$$

which is equation (79).

D.5.3 Solution for aggregate output Y_t^{os} .

From A.63, $Y_t = N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}}$. Then: v_{os}

$$\begin{aligned} Y_t^{os} &= N_t^{\frac{\sigma}{\sigma-1} - \frac{1}{1-\eta}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} L_{pt}^{\frac{1}{1-\eta}} \\ &= N_t^{\frac{\sigma}{\sigma-1}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} v_{os}^{\frac{1}{1-\eta}} \\ &= (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} v_{os} \\ &= (v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}} \end{aligned}$$

and this is equation (82).

D.5.4 Consumption per capita and growth

$$c_t^{os} = \frac{Y_t^{os}}{L_t} \propto L_t^{\frac{1}{\sigma-1}}$$

which is equation (84). Thus, consumption per capita growth is:

$$g_c^{os} = \left(\frac{1}{\sigma-1} \right) g_L$$

which is is equation (86).

D.5.5 Firm Production Y_{it}^{os}

Using equation (82) of the paper:

$$\begin{aligned}
Y_{it}^{os} &= Y_t N_t^{-\frac{\sigma}{\sigma-1}} \\
&= (v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}} N_t^{-\frac{\sigma}{\sigma-1}} \\
&= (v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}} (L_t \psi_{os})^{-\frac{\sigma}{\sigma-1}} \\
&= (v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}}
\end{aligned}$$

which is equation (92).

D.5.6 Data used by the firm D_{it}^{os}

By definition, $D_{it} = \alpha x_{it} Y_{it}$. Then:

$$\begin{aligned}
D_{it}^f &= \alpha x_{it} Y_{it} \\
&= (v_{os} \alpha \bar{x})^{\frac{1}{1-\eta}}
\end{aligned}$$

which is equation (88).

D.5.7 Aggregate Data used by the firm D_t^{os}

By definition $D_t = N_t D_{it}$, hence:

$$\begin{aligned}
D_t^f &= N_t (v_{os} \alpha \bar{x})^{\frac{1}{1-\eta}} \\
&= (v_{os} \alpha \bar{x})^{\frac{1}{1-\eta}} L_t \psi_{os}
\end{aligned}$$

which is equation (90).

D.5.8 Labor share $\left(\frac{w_t L_{pt}}{Y_t} \right)$

From A.64:

$$\begin{aligned}
\left(\frac{w_t L_{pt}}{Y_t}\right)^{os} &= \frac{L_{pt}}{Y_t} \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} \\
&= ((v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}})^{-1} \frac{(\alpha \bar{x})^{\frac{\eta}{1-\eta}}}{1-\eta} \left(1 - \frac{1}{\sigma}\right) L_{pt}^{\frac{\eta}{1-\eta} + 1} N_t^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} \\
&= ((v_{os})^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}})^{-1} \frac{1}{1-\eta} \left(1 - \frac{1}{\sigma}\right) (v_{os} L_t \psi_{os})^{\frac{\eta}{1-\eta} + 1} (L_t \psi_{os})^{\frac{1}{\sigma-1} - \frac{\eta}{1-\eta}} \\
&= \frac{1}{1-\eta} \left(1 - \frac{1}{\sigma}\right) \\
&= \frac{1-\sigma}{\sigma(1-\eta)}
\end{aligned}$$

which is equation (93).

D.5.9 Profit share $\left(\frac{N_t \pi_t}{Y_t}\right)$

From A.65, $\pi_t = L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]$. Hence:

$$\begin{aligned}
\left(\frac{\pi_t N_t}{Y_t}\right)^{os} &= \frac{N_t L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta}} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{Y_t} \\
&= \frac{L_{pt}^{\frac{1}{1-\eta}} N_t^{\frac{1}{\sigma-1} - \frac{1}{1-\eta} + 1} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{(v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}}} \\
&= \frac{(v_{os} L_t \psi_{os})^{\frac{1}{1-\eta}} (L_t \psi_{os})^{\frac{1}{\sigma-1} - \frac{1}{1-\eta} + 1} (\alpha \bar{x})^{\frac{\eta}{1-\eta}} \left[1 - \frac{\sigma-1}{\sigma(1-\eta)}\right]}{(v_{os} \alpha^\eta \bar{x}^\eta)^{\frac{1}{1-\eta}} (\psi_{os} L_t)^{\frac{\sigma}{\sigma-1}}} \\
&= 1 - \frac{\sigma-1}{\sigma(1-\eta)} \\
&= \frac{1-\eta\sigma}{\sigma(1-\eta)}
\end{aligned}$$

and this is equation (94).

D.5.10 Price of a variety p_{it}^{os}

From the household problem, $p_{it} = \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}}$, then:

$$\begin{aligned} p_{it}^{os} &= \left(\frac{Y_t}{Y_{it}}\right)^{\frac{1}{\sigma}} \\ &= \left(\frac{Y_t}{Y_t N_t^{-\frac{\sigma}{\sigma-1}}}\right)^{\frac{1}{\sigma}} \\ &= N_t^{\frac{1}{\sigma-1}} \end{aligned}$$

and from equation (79), $N_t = \psi_{os} L_t$, thus:

$$p_{it}^f = (\psi_{os} L_t)^{\frac{1}{\sigma-1}}$$

which is equation (99).