ECON272A - Problem Set #1

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1.1 Conceptual Questions

Question 1

Assumption	Example Question	Ruled out because
A1(a)	Does the elasticity of output with re-	Output elasticities (β_k, β_l) are con-
	spect to inputs (capital and labor)	stant for all levels of output.
	change as output increases?	
A1(b)	Are there fixed material costs in-	Materials are in fixed proportion to
	curred from production?	capital and labor.
A1(c)	How does investment risk affect the	Capital accumulation is determinis-
	capital accumulation of the firm?	tic.
A1(d)	Can firm investments lead to	Productivity is modeled as an AR(1)
	endogenous productivity improve-	process; ω_{it} depends only on $\omega_{i,t-1}$.
	ments in the same period?	
A1(e)	Is productivity growth factor-	$\omega_{it} \in \mathbb{R}$ and only captures factor-
	augmenting for the firm?	neutral productivity.

Question 2

Assumption	Example Question	Ruled out because
A2(a)	Do firms produce differentiated va-	All firms produce the same output
	rieties?	and belong to the same market.
A2(b)	Do different firms have different val-	Firms discount the future with a
	ues for future profits?	common factor $(1-r)$.
A2(c)	Does the firm have monopsony	Wages are determined by an $AR(1)$
	power in the labor market?	process.
A2(d)	How do changes in investment cost	The cost of investment varies only
	over time affect firms?	across firms but not over time.

Question 3

- Assumptions A1(b),(e) imply that productivity shocks are Hicks-neutral.
- Assumptions A1(d),(e) imply that the correlation coefficient ρ in (2) is a scalar value.
- Assumptions A2(a),(b) imply that any two or more firms with the same state variables (k_{it}, ω_{it}) will face the same profit maximization problem and behave identically.
- Assumption A2(d) implies that the firm chooses the same level of optimal investment ($I_{it} = I^*$) in every period. This, with Assumption A1(c), implies one of the following 3 cases for the capital stock of the firm: (i) K grows at a constant rate (if $I^* > \delta K_{it}$), (ii) K has a steady-state (if $I^* = \delta K_{it}$) or (iii) K eventually falls to 0 (if $I^* < \delta K_{it}$).

Question 4

Equation (2) imposes the structural assumption that $\omega_{it} = \rho \omega_{i,t-1} + \xi_{it}$ where ξ_{it} is a mean zero i.i.d. random shock.

- To justify the use of OLS to estimate (1), we have to assume that $\omega_{it} = 0$ for all i and for all t. This is equivalent to assuming $\rho = 0$ in (2).
- To justify the use of FE to estimate (1), we have to assume that $\omega_{it} = \omega_i$ for all t. This is equivalent to assuming $\rho = 1$ in (2).

Therefore, the assumptions required for OLS and FE can be interpreted as special cases of the model that specify the value of ρ in equation (2).

1.2 Simulation

Questions 1, 2(d) to 2(g), 3

See below for code.

Question 2a

Let $A_{it} = \exp(\beta_0) \exp(\omega_{it}) \exp(\varepsilon_{it})$. The production function is Cobb-Douglas in capital and labor, and can be written as:

$$Q_{it} = F(K_{it}, L_{it}) = A_{it} K_{it}^{\beta_k} L_{it}^{\beta_l}$$

Which gives the first-order condition (FOC) with respect to labor:

$$P_{it}^l = A_{it} K_{it}^{\beta_k} \beta_l L_{it}^{\beta_l - 1}$$

Re-arranging, the optimal labor input is:

$$L_{it} = \left(\frac{A_{it} K_{it}^{\beta_k} \beta_l}{P_{it}^l}\right)^{\frac{1}{1-\beta_l}}$$

Question 2b

Capital evolves according to:

$$K_{i,t+1} = (1 - \delta)K_{it} + I_{it}$$

with convex investment costs:

$$C(I_{it}; \gamma_i) = \frac{1}{2} \gamma_i I_{it}^2$$

Next-period production can thus be expressed as:

$$Q_{i,t+1} = A_{i,t+1} K_{i,t+1}^{\beta_k} L_{i,t+1}^{\beta_l} = A_{i,t+1} [(1-\delta)K_{it} + I_{it}]^{\beta_k} L_{i,t+1}^{\beta_l}$$

Using the labor FOC, the FOC with respect to investment-today:

$$\gamma_{i}I_{it} = A_{i,t+1}L_{i,t+1}^{\beta_{l}}\beta_{k}[(1-\delta)K_{it} + I_{it}]^{\beta_{k}-1}$$

$$= A_{i,t+1}\left(\frac{A_{i,t+1}K_{i,t+1}^{\beta_{k}}\beta_{l}}{P_{i,t+1}^{l}}\right)^{\frac{\beta_{l}}{1-\beta_{l}}}\beta_{k}[K_{i,t+1}]^{\beta_{k}-1}$$

$$= A_{i,t+1}\left(\frac{A_{i,t+1}\beta_{l}}{P_{i,t+1}^{l}}\right)^{\frac{\beta_{l}}{1-\beta_{l}}}\beta_{k}[K_{i,t+1}]^{\frac{\beta_{k}\beta_{l}}{1-\beta_{l}}+\beta_{k}-1}$$

$$= A_{i,t+1}\left(\frac{A_{i,t+1}\beta_{l}}{P_{i,t+1}^{l}}\right)^{\frac{\beta_{l}}{1-\beta_{l}}}\beta_{k}$$

which is independent of the level of capital because if $\beta_k + \beta_l = 1$, then

$$\frac{\beta_k \beta_l}{1 - \beta_l} + \beta_k - 1 = \frac{\beta_k \beta_l + \beta_k (1 - \beta_l) - 1 + \beta_l}{1 - \beta_l} = \frac{(\beta_k + \beta_l) - 1}{1 - \beta_l} = 0$$

Question 2c

Given that $\beta_0 = 0$ and $\mathbb{E}[\exp(\varepsilon_{it})] = 1$,

$$\mathbb{E}[A_{it}] = \mathbb{E}[\exp(0)\exp(\omega_{it})\exp(\varepsilon_{it})] = \exp(\omega_{it})\mathbb{E}[\exp(\varepsilon_{it})] = \exp(\omega_{it})$$

Where $\exp(\omega_{it})$ is outside the expectation operator because of Assumption A2(b) (each firm takes as given their ω_{it} in each period t). Now we can write the optimal investment rule as:

$$I_{it} = \frac{1}{\gamma_i} \mathbb{E}[A_{i,t+1}] \left(\frac{\mathbb{E}[A_{i,t+1}]\beta_l}{P_{i,t+1}^l}\right)^{\frac{\beta_l}{1-\beta_l}} \beta_k$$
$$= \frac{1}{\gamma_i} \exp(\omega_{i,t+1}) \left(\frac{\exp(\omega_{i,t+1})\beta_l}{P_{i,t+1}^l}\right)^{\frac{\beta_l}{1-\beta_l}} \beta_k$$

The investment policy equation is the present discounted value of future investment flows (net of depreciation) at t = 0:

$$\kappa(\{\omega_{it}, P_{it}^l\}_{t=1}^{\infty}) = \sum_{t=1}^{\infty} (1 - \delta)^t (1 - r)^t I_{it}$$

$$= \sum_{t=1}^{\infty} (1 - \delta)^t (1 - r)^t \frac{1}{\gamma_i} \exp(\omega_{it}) \left(\frac{\exp(\omega_{it})\beta_l}{P_{it}^l}\right)^{\frac{\beta_l}{1 - \beta_l}} \beta_k$$

$$= \frac{1}{\gamma_i} \cdot \beta_k \sum_{t=1}^{\infty} (1 - \delta)^t (1 - r)^t \exp(\omega_{it}) \left(\frac{\exp(\omega_{it})\beta_l}{P_{it}^l}\right)^{\frac{\beta_l}{1 - \beta_l}}$$

Python code for '1.2 Simulation'

```
import pandas as pd
import numpy as np
# (1a) Set seed and parameters
np.random.seed(123456789)
N=1000
T = 50
beta_zero=0
beta_k=0.4
beta_1=0.6
rho=0.7
theta=0.3
delta=0.2
r = 0.05
# (1b) Initial conditions for productivity and wages
sigma_omega=np.sqrt(0.3)
sigma_logwage=np.sqrt(0.1)
omega_zero=np.random.normal(0,sigma_omega,(N,1)) # omeqa_0 \circ N(0,0.3)
wage_zero=np.exp(np.random.normal(0,sigma_logwage,(N,1))) # log wage_0 ~ N(0,0.1)
# (1b) Shocks to productivity and wages (i.i.d. across firms and time)
sigma_xi=np.sqrt(0.214)
sigma_nu=np.sqrt(0.1)
xi=np.random.normal(0,sigma_xi,(N,T)) # xi ~ N(0,0.214)
nu=np.exp(np.random.normal(0,sigma_nu,(N,T))) # log nu ~ N(0,0.1)
# (1c) Paths for productivity and wages
M=np.zeros((N,T))
omega=np.concatenate((omega_zero,M),axis=1)
wage=np.concatenate((wage_zero,M),axis=1)
for t in range(T):
        omega[:,t+1]=rho*omega[:,t]+xi[:,t] # Productivity path from t=0 to t=50
        wage[:,t+1]=theta*wage[:,t]+nu[:,t] # Wage path from t=0 to t=50
# (1c) Generate time-invariant investment costs for firms
sigma_g=np.sqrt(0.6)
g=np.random.normal(0,sigma_g,(N,1)) # g = log(1/gamma)^N(0,0.6)
gamma=1/np.exp(g) # gamma = 1/exp(g)
# (2d) Path for investments from t=0 to t=49 (use FOC for investments)
inv=np.zeros((N,T))
for i in range(N):
        for t in range(T):
                inv[i,t]=1/gamma[i]*beta_k*np.exp(omega[i,t+1])*\
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np.power((np.exp(omega[i,t+1])*beta_1/wage[i,t+1]),(beta_1/(1-beta_1)))
# (2d) Path for capital from t=0 to t=50 (use equation for capital growth)
# Assume that every firm begins with 1 unit of capital at t=0
capital=np.ones((N,T+1))
for t in range(T):
        capital[:,t+1]=(1-delta)*capital[:,t]+inv[:,t]
log_capital=np.log(capital)
# (2e) Path for labor from t=0 to t=50 (use FOC for labor)
labor=np.power((np.exp(omega)*np.power(capital,beta_k)*beta_l)/wage,1/(1-beta_l))
log_labor=np.log(labor)
# (2f) Simulate measurement error
sigma_err=np.sqrt(0.1)
err=np.random.normal(0,sigma_err,(N,T+1))
# (2f) Path for output (use equation 1)
log_output=beta_zero+beta_k*log_capital+beta_l*log_labor+omega+err
output=np.exp(log_output)
# (2q) Path for materials (output prior to measurement error)
log_materials=log_output-omega-err
materials=np.exp(log_materials)
# (3) Dataframe for log_output (q_it), log_capital (k_it),
# log_labor (l_it), log_materials (m_it)
# (3) Index for N firms
firms=pd.DataFrame(np.linspace(1,N,num=N), columns=["firm_id"])
firm_id=pd.concat([firms]*51,ignore_index=True)
# (3) Index for T+1 periods
time=pd.Series(np.linspace(0,T,num=T+1)).repeat(1000)
time=pd.DataFrame(time, columns=["time"]).reset_index()
# (3) Reshape 'N x (T+1)' matrices into columns of length N*(T+1)
q_it=pd.DataFrame(log_output.flatten('F'), columns=["log_output"])
k_it=pd.DataFrame(log_capital.flatten('F'), columns=["log_capital"])
l_it=pd.DataFrame(log_labor.flatten('F'), columns=["log_labor"])
m_it=pd.DataFrame(log_materials.flatten('F'), columns=["log_materials"])
# (3) Assemble dataframe
dataframe=firm_id.join(time['time']).join(q_it).join(k_it).join(l_it).join(m_it)
```