

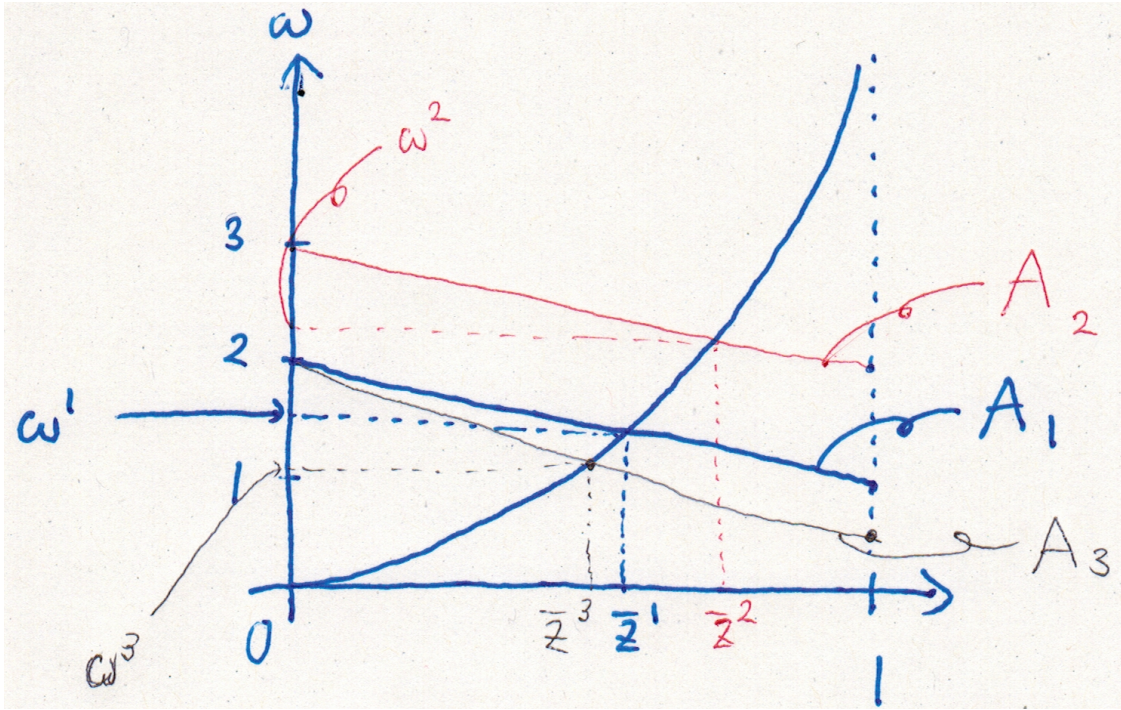
# Term Paper: Answer Key

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## 1 The Dornbusch-Fischer-Samuelson model

Figure 1: Comparative statics with respect to  $A$



Recall that the equilibrium relative wage  $\bar{\omega}$  and the equilibrium cutoff (of goods)  $\bar{z}$  are characterized as the intersection of the downward-sloping  $A$  curve and the upward-sloping  $B$  curve.  $B$  doesn't change throughout this exercise.

Initially, the relative unit labor requirement  $A$  is  $A_1 = 2 - z$ . Now suppose that this shifts up to  $A_2 = 3 - z$ . Remember that  $A(z) = a^*(z)/a(z)$ , where  $a^*(z)$  and  $a(z)$  are Foreign and Home's unit labor requirement to produce good  $z$ , respectively. Therefore, for all goods, Foreign becomes less productive. Then the cutoff shifts from  $\bar{z}^1$  to  $\bar{z}^2$ . Recall that the cutoff  $\bar{z}$  is such that Home produces goods  $[0, \bar{z}]$  and Foreign produces goods  $[\bar{z}, 1]$ . Therefore, because of this shift of relative unit labor requirement, Home produces a wider range of goods and Foreign produces a narrower range of goods. Accordingly, the relative wage between Home and Foreign increases from  $\omega^1$  to  $\omega^2$ . In short, Home becomes productive uniformly for all goods; Home becomes an exporter of a wider range of goods; and the wage ratio between Home and Foreign increases.

Now the relative unit labor requirement rotates clockwise from  $A_1$  to  $A_2$ . For any  $z \in (0, 1]$ , the relative unit labor requirement decreases. This means that for all goods except good 0, Home becomes relatively less

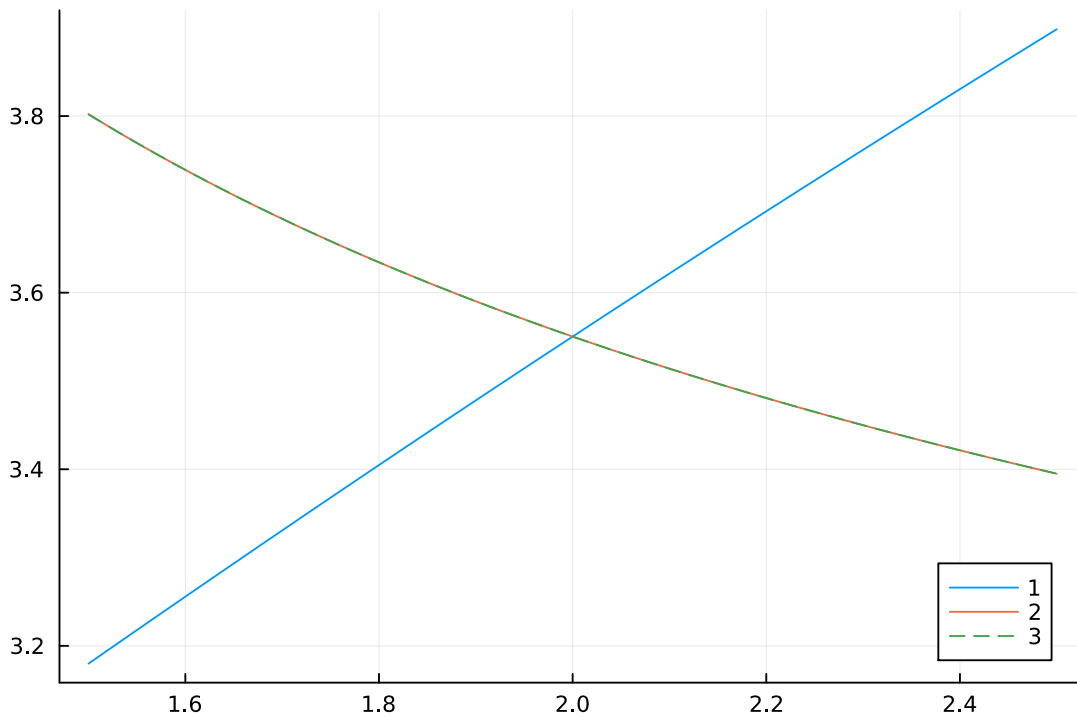
productive. Therefore, the cutoff decreases from  $\bar{z}^1$  to  $\bar{z}^3$ . In short, Home becomes less productive for almost all goods; Home becomes an importer of a wider range of goods; and the relative wage between Home and Foreign decreases.

## 2 Computing the Eaton-Kortum model

### 2.1 Comparative statics with respect to $T_1$

1. See Figure 2.

Figure 2: Q2.1.1



2. See Figure 3.

### 2.2 Comparative statics with respect to $d_{12} = d_{21}$

1. See Figure 4.
2. See Figure 5.
3. See Figure 6.

Figure 3: Q2.1.2

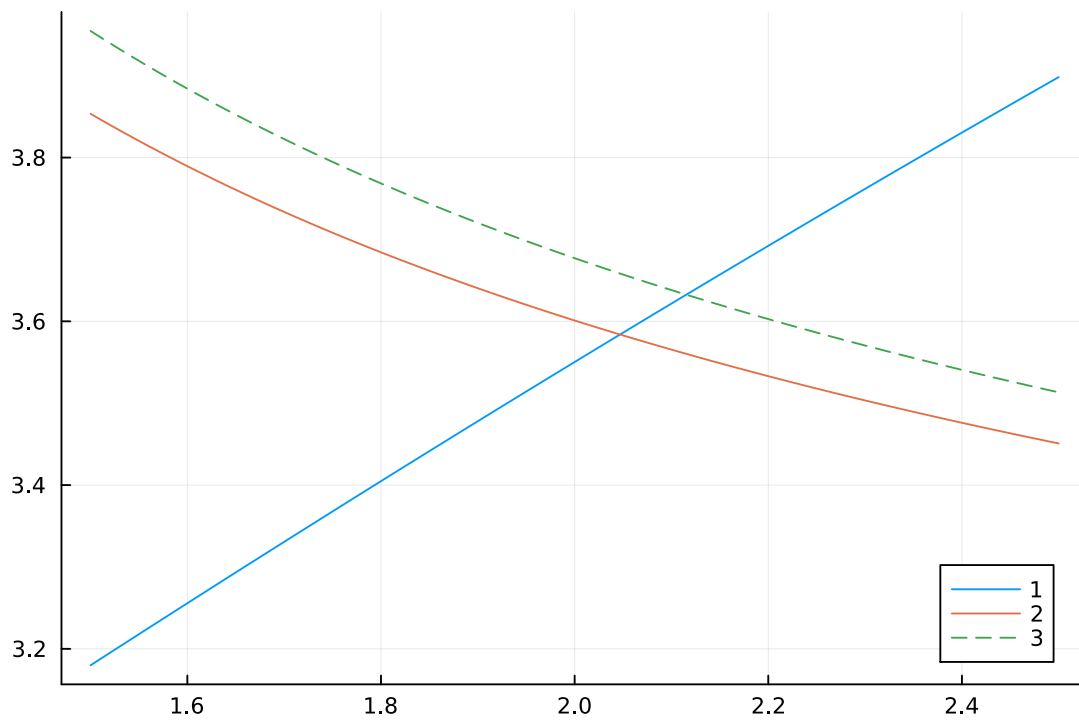


Figure 4: Q2.2.1

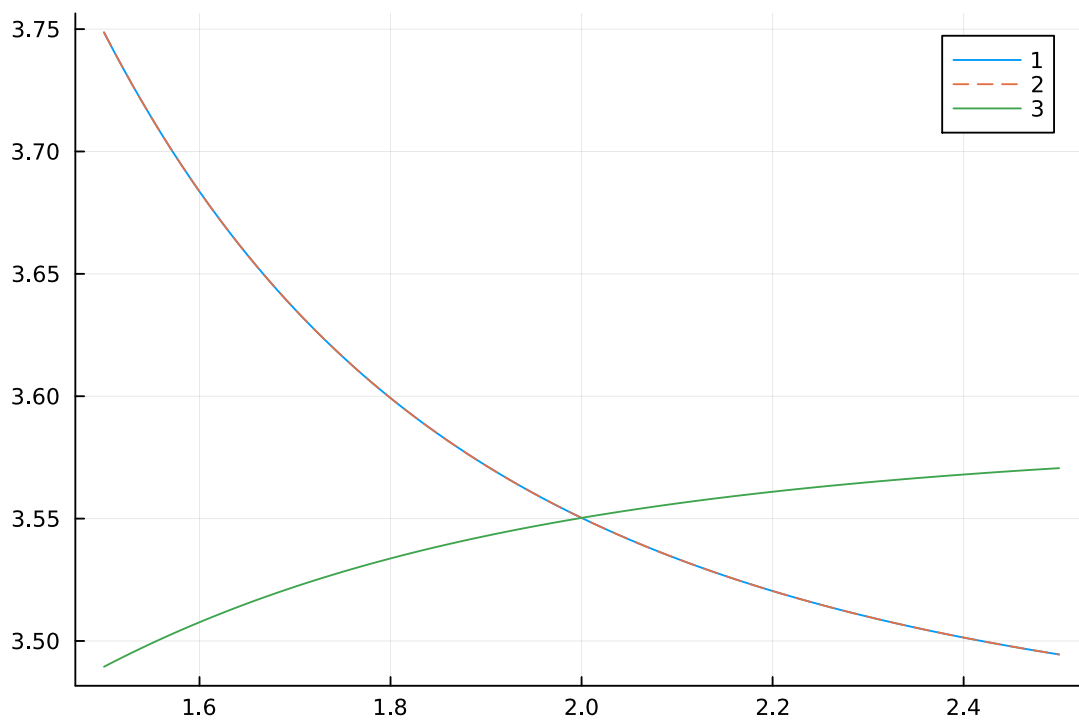


Figure 5: Q2.2.2

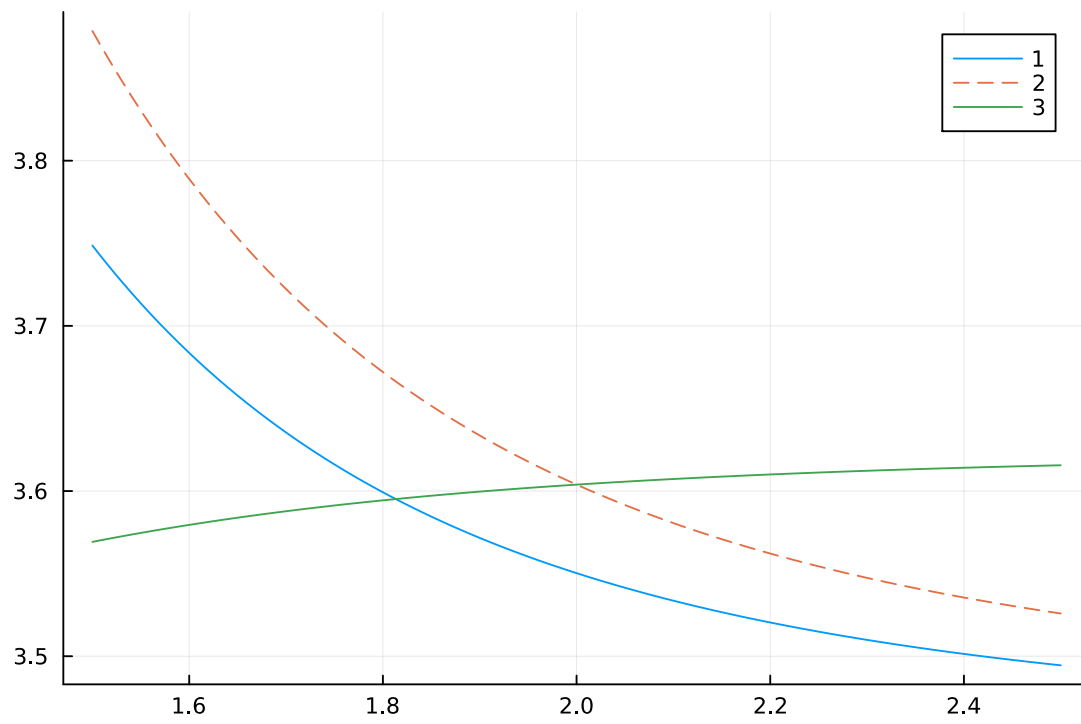
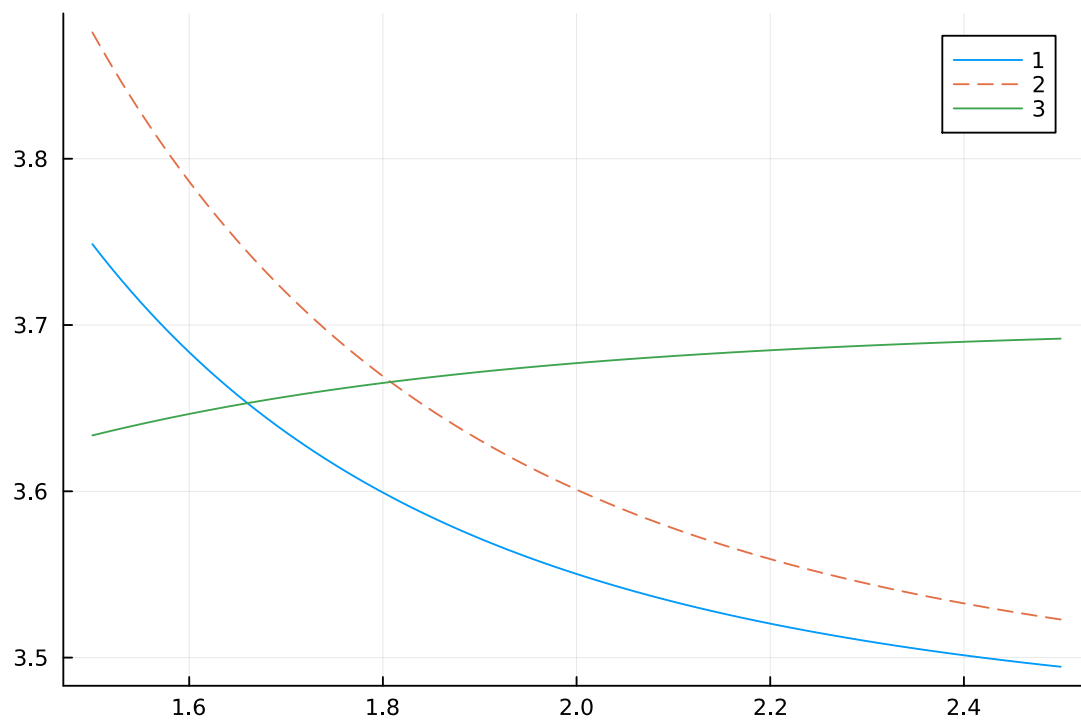


Figure 6: Q2.2.3



### 3 The Melitz-Chaney model

1. The firm's profit-maximization problem is

$$\max_{\{p_{ij}(\varphi)\}_{j \in S}} \sum_{j \in S} \left( p_{ij}(\varphi)^{1-\sigma} Y_j P_j^{\sigma-1} - \frac{w_i}{\varphi} \tau_{ij} p_{ij}(\varphi)^{-\sigma} Y_j P_j^{\sigma-1} - f_{ij} \right).$$

The first-order condition with respect to  $p_{ij}(\varphi)$  is

$$(1 - \sigma) p_{ij}(\varphi)^{-\sigma} Y_j P_j^{\sigma-1} + \sigma \frac{w_i}{\varphi} \tau_{ij} p_{ij}(\varphi)^{-\sigma-1} Y_j P_j^{\sigma-1} = 0.$$

Rearranging this, we have

$$\sigma \frac{w_i}{\varphi} \tau_{ij} p_{ij}(\varphi)^{-\sigma-1} = (\sigma - 1) p_{ij}(\varphi)^{-\sigma}.$$

Solving this for  $p_{ij}(\varphi)$ ,

$$p_{ij}(\varphi) = \frac{\sigma}{\sigma - 1} \frac{w_i}{\varphi} \tau_{ij}.$$

2. Our starting point is

$$f_i^e = \sum_{j \in S} \int_{\varphi_{ij}^*}^{\infty} (\pi_{ij}(\varphi) - f_{ij}) dG_i(\varphi).$$

Substituting  $\pi_{ij}(\varphi) = \frac{1}{\sigma} \left( \frac{\sigma}{\sigma-1} \frac{w_i}{\varphi} \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1}$  and  $g_i(\varphi) = \theta_i \varphi^{-(\theta_i+1)}$  into this, we have

$$\begin{aligned} f_i^e &= \sum_{j \in S} \int_{\varphi_{ij}^*}^{\infty} \left[ \frac{1}{\sigma} \left( \frac{\sigma}{\sigma-1} \frac{w_i}{\varphi} \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} - f_{ij} \right] \theta_i \varphi^{-(\theta_i+1)} d\varphi \\ &= \sum_{j \in S} \int_{\varphi_{ij}^*}^{\infty} \left[ \frac{\theta_i}{\sigma} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} \varphi^{\sigma-2-\theta_i} - \theta_i f_{ij} \varphi^{-(\theta_i+1)} \right] d\varphi \\ &= \sum_{j \in S} \left[ \frac{1}{\sigma - \theta_i - 1} \frac{\theta_i}{\sigma} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} \varphi^{\sigma-\theta_i-1} + \frac{1}{\theta_i} \theta_i f_{ij} \varphi^{-\theta_i} \right]_{\varphi_{ij}^*}^{\infty}. \end{aligned}$$

Since  $\theta_i > \sigma - 1$ , we have  $\sigma - \theta_i - 1 < 0$ . Of course we have  $-\theta_i < 0$ . Therefore, we have

$$f_i^e = \sum_{j \in S} \left[ -\frac{1}{\sigma - \theta_i - 1} \frac{\theta_i}{\sigma} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} (\varphi_{ij}^*)^{\sigma-\theta_i-1} - f_{ij} (\varphi_{ij}^*)^{-\theta_i} \right].$$

Substituting  $\varphi_{ij}^* = \left( \frac{\sigma f_{ij} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{\sigma-1}}{Y_j P_j^{\sigma-1}} \right)^{\frac{1}{\sigma-1}}$  into this,

$$f_i^e = \sum_{j \in S} \left[ \frac{1}{\theta_i + 1 - \sigma} \frac{\theta_i}{\sigma} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} \left( \frac{\sigma f_{ij} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{\sigma-1}}{Y_j P_j^{\sigma-1}} \right)^{\frac{\sigma-\theta_i-1}{\sigma-1}} - f_{ij} \left( \frac{\sigma f_{ij} \left( \frac{\sigma}{\sigma-1} w_i \tau_{ij} \right)^{\sigma-1}}{Y_j P_j^{\sigma-1}} \right)^{-\frac{\theta_i}{\sigma-1}} \right]. \quad (1)$$

The first term in the square bracket is

$$\begin{aligned} & \frac{1}{\theta_i + 1 - \sigma} \frac{\theta_i}{\sigma} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{1-\sigma} Y_j P_j^{\sigma-1} (\sigma f_{ij})^{1-\frac{\theta_i}{\sigma-1}} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{\sigma-\theta_i-1} (Y_j P_j^{\sigma-1})^{-1+\frac{\theta_i}{\sigma-1}} \\ &= \frac{\theta_i}{\theta_i + 1 - \sigma} \sigma^{-\frac{\theta_i}{\sigma-1}} f_{ij}^{1-\frac{\theta_i}{\sigma-1}} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{-\theta_i} (Y_j P_j^{\sigma-1})^{\frac{\theta_i}{\sigma-1}}. \end{aligned}$$

The second term in the square bracket is

$$(f_{ij})^{1-\frac{\theta_i}{\sigma-1}} \sigma^{-\frac{\theta_i}{\sigma-1}} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{-\theta_i} (Y_j P_j^{\sigma-1})^{\frac{\theta_i}{\sigma-1}}.$$

Then the right-hand side of (1) is

$$\begin{aligned} & \sum_{j \in S} \left[ \frac{\theta_i}{\theta_i + 1 - \sigma} \sigma^{-\frac{\theta_i}{\sigma-1}} f_{ij}^{1-\frac{\theta_i}{\sigma-1}} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{-\theta_i} (Y_j P_j^{\sigma-1})^{\frac{\theta_i}{\sigma-1}} - (f_{ij})^{1-\frac{\theta_i}{\sigma-1}} \sigma^{-\frac{\theta_i}{\sigma-1}} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{-\theta_i} (Y_j P_j^{\sigma-1})^{\frac{\theta_i}{\sigma-1}} \right] \\ &= \sum_{j \in S} \frac{\sigma - 1}{\theta_i + 1 - \sigma} \left( \frac{\sigma}{\sigma - 1} w_i \tau_{ij} \right)^{-\theta_i} \sigma^{-\frac{\theta_i}{\sigma-1}} f_{ij}^{\frac{\sigma-\theta_i-1}{\sigma-1}} Y_j^{\frac{\theta_i}{\sigma-1}} P_j^{\theta_i}. \end{aligned}$$

## 4 Production networks

1. The cost minimization problem is

$$\begin{aligned} & \min_{L_i, X_j} L_i + \sum_{j=1}^n P_j X_{ij}, \\ & \text{s.t.} \quad e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{\sum_{j=1}^n 1-\alpha_{ij}} \prod_{j=1}^n X_{ij}^{\alpha_{ij}} \geq 1. \end{aligned} \tag{2}$$

Set up the Lagrangian

$$\begin{aligned} \mathcal{L} &= L_i + \sum_{j=1}^n P_j X_{ij} \\ &+ \lambda \left( e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{\sum_{j=1}^n 1-\alpha_{ij}} \prod_{j=1}^n X_{ij}^{\alpha_{ij}} - 1 \right). \end{aligned}$$

The first-order conditions are

$$\frac{\partial \mathcal{L}}{\partial L_i} = 1 + \lambda \left( 1 - \sum_{j=1}^n \alpha_{ij} \right) e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{-\sum_{j=1}^n \alpha_{ij}} \prod_{j=1}^n X_{ij}^{\alpha_{ij}} = 0, \tag{3}$$

$$\frac{\partial \mathcal{L}}{\partial X_{ij}} = P_j + \lambda \alpha_{ij} e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{1-\sum_{j=1}^n \alpha_{ij}} \left( \prod_{j' \neq j} X_{ij'}^{\alpha_{ij'}} \right) X_{ij}^{\alpha_{ij}-1} = 0 \tag{4}$$

for any  $j = 1, \dots, n$ , and

$$\frac{\partial \mathcal{L}}{\partial \lambda} = e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{1 - \sum_{j=1}^n \alpha_{ij}} \prod_{j=1}^n X_{ij}^{\alpha_{ij}} - 1 = 0. \quad (5)$$

Rearranging (3) and (4) and taking the ratio of these two yields

$$\frac{1}{P_j} = \left( \frac{1 - \sum_{j'=1}^n \alpha_{ij'}}{\alpha_{ij}} \right) \cdot \frac{X_{ij}}{L_i}.$$

Solving this for  $X_{ij}$ ,

$$X_{ij} = \frac{1}{P_j} \left( \frac{\alpha_{ij}}{1 - \sum_{j'=1}^n \alpha_{ij'}} \right) L_i \quad (6)$$

for any  $j$ . Substituting this into (5) yields

$$e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i^{1 - \sum_{j=1}^n \alpha_{ij}} \cdot \prod_{j=1}^n \left[ \frac{1}{P_j} \left( \frac{\alpha_{ij}}{1 - \sum_{j'=1}^n \alpha_{ij'}} \right) L_i \right]^{\alpha_{ij}} = 1.$$

Rearranging this,

$$e^{\varepsilon_i} A_i(\alpha_i) \zeta(\alpha_i) L_i \cdot \left( \prod_{j=1}^n \left( \frac{1}{P_j} \right)^{\alpha_{ij}} \right) \cdot \left( \prod_{j=1}^n \alpha_{ij}^{\alpha_{ij}} \right) \cdot \left( \prod_{j=1}^n \left( \frac{1}{1 - \sum_{j'=1}^n \alpha_{ij'}} \right)^{\alpha_{ij}} \right) = 1.$$

Using the definition of  $\zeta(\alpha_i)$  (See p.8 of `7_production_network.pdf`),

$$e^{\varepsilon_i} A_i(\alpha_i) L_i \cdot \left( \prod_{j=1}^n \left( \frac{1}{P_j} \right)^{\alpha_{ij}} \right) \cdot \left( 1 - \sum_{j=1}^n \alpha_{ij} \right)^{-1} = 1.$$

Solving this for  $L_i$ ,

$$L_i = \left( 1 - \sum_{j=1}^n \alpha_{ij} \right) \cdot \prod_{j=1}^n P_j^{\alpha_{ij}} \cdot \frac{1}{e^{\varepsilon_i} A_i(\alpha_i)}. \quad (7)$$

Substituting this into (6) yields

$$X_j = \frac{\alpha_{ij}}{P_j} \cdot \left( \prod_{j'=1}^n P_{j'}^{\alpha_{ij'}} \right) \cdot \frac{1}{e^{\varepsilon_i} A_i(\alpha_i)}. \quad (8)$$

These are the cost-minimizing labor/intermediate inputs. Plugging (7) & (8) into the objective function

(2):

$$\begin{aligned} & \left(1 - \sum_{j'=1}^n \alpha_{ij'}\right) \cdot \prod_{j=1}^n P_j^{\alpha_{ij}} \cdot \frac{1}{e^{\varepsilon_i} A_i(\alpha_i)} + \sum_{j'=1}^n P_{j'} \cdot \frac{\alpha_{ij'}}{P_{j'}} \cdot \left(\prod_{j=1}^n P_j^{\alpha_{ij}}\right) \cdot \frac{1}{e^{\varepsilon_i} A_i(\alpha_i)} \\ &= \frac{\prod_{j=1}^n P_j^{\alpha_{ij}}}{e^{\varepsilon_i} A_i(\alpha_i)}. \end{aligned}$$

2. To compute the Hessian, first we compute the first derivatives.

$$\frac{\partial a_2}{\partial \alpha_{21}} = -2(\alpha_{21} - 0.4) - 2(\alpha_{21} - \alpha_{22} - 0.1),$$

$$\frac{\partial a_2}{\partial \alpha_{22}} = -2(\alpha_{22} - 0.4) + 2(\alpha_{21} - \alpha_{22} - 0.1).$$

The second derivatives are

$$\frac{\partial^2 a_2}{\partial \alpha_{21}^2} = -4,$$

$$\frac{\partial^2 a_2}{\partial \alpha_{22}^2} = -4,$$

and because of Young's theorem, we have

$$\frac{\partial^2 a_2}{\partial \alpha_{21} \partial \alpha_{22}} = \frac{\partial^2 a_2}{\partial \alpha_{22} \partial \alpha_{21}} = -2.$$

Therefore, the Hessian is

$$\begin{bmatrix} -4 & 2 \\ 2 & -4 \end{bmatrix}.$$

Its inverse matrix is

$$\frac{1}{16-4} \begin{bmatrix} -4 & -2 \\ -4 & -2 \end{bmatrix} = \begin{bmatrix} -1/3 & -1/6 \\ -1/6 & -1/3 \end{bmatrix}.$$

The off-diagonal elements are -1/6. Inputs 1 and 2 are complements for sector 2.