Comments Panel on Open Questions

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Conference on Matching and Price Theory

Milton Friedman Institute University of Chicago May 7, 2011

Bounds

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- Let A be square (not really needed). See C. R. Rao (1971) λ is matrix of eigenvalues of A. Then we know from linear

algebra that there exists $\begin{array}{cc} P & (M & r) & \lambda(r & r) \\ Q & M & r & A = P\lambda Q' \end{array}$ where columns of P are orthonormal (mutually orthogonal L unit length)

true even if A is m n $m \notin n$ P is m r λ is r r Q is n r.

• Unique if all $\lambda_i > 0$. Then

$$a_{ij} = \sum_{k=1}^r \lambda_k P_{ik} q_{jk}.$$

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• The spectral decomposition assigns a Cobb-Douglas interaction to each component:

 P_{ik} is quality k of worker i q_{jk} is quality k of firm j.

• ... implicitly we have a Cobb-Douglas technology in qualities

$$\begin{pmatrix} P_{11} & P_{1r} \\ \vdots & \vdots \\ P_{n1} & P_{nr} \end{pmatrix} \begin{pmatrix} \lambda_1 & 0 \\ & \ddots & \\ 0 & \lambda_r \end{pmatrix}$$
$$\begin{pmatrix} q_{11} & q_{n1} \\ \vdots & \vdots \\ q_{1r} & q_{nr} \end{pmatrix}.$$

• Observe $a_{ii} = w_i + \pi_i$. Now $a_{ji} = w_j + \pi_i$

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 $a_{ji} a_{ii} w_j w_i.$

• \therefore W_i W_j a_{ii} a_{ji} .

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- ∴a_{ij} a_{ii} w_i w_j a_{ii} a_{ji}
- Now use spectral decomposition

$$\begin{array}{lll} \textbf{a}_{ij} & \textbf{a}_{jj} & = & \sum\limits_{k=1}^{r} \lambda_k \begin{pmatrix} P_{ik} & P_{jk} \end{pmatrix} q_{jk} \\ & \text{like marginal product} \\ & \sum\limits_{k=1}^{r} \lambda_k (P_{ik} & P_{jk}) q_{jk} & w_i & w_j & \sum\limits_{k=1}^{r} \lambda_k (P_{ik} & P_{jk}) q_{ik} \end{array}$$

• Left and right hand sides are differential marginal product using firm j and firm i^s attributes, respectively. Suppose all firms alike: $q_{jk} = q_{ik}$ (firms possess no identity). Then we get Gorman-Lancaster form of the model

$$w_i \quad w_j = \sum_{k=1}^r \lambda_k (P_{ik} \quad P_{jk}) q_k$$

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• Otherwise, we get the notion that workers have different productivities depending on properties of firms). (Then characteristics payment will depend on distributions of firms.

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- Agents cannot invest in order to change skills s.
- $t_i(s)$ is a function that expresses the amount of sector *i* specific tasks a worker with endowment of skills *s* can perform.

Roy Model	Estimates
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- π_i is the price of one unit of sector *i* specific task.

Estimates

Optimization implies:

(1)
$$\pi_i = P_i \frac{\partial F^{(i)}}{\partial T_i}$$

An agent with endowment s works in sector i if:

(2)
$$\pi_i t_i(s) \quad \pi_j t_j(s), \ i, j = 1, 2 \text{ and } i \notin j.$$

Let L_i denote the set of agents working in sector i:

$$L_{i} = fs: \pi_{i}t_{i}(s) \quad \pi_{j}t_{j}(s), i \notin jg.$$

The log wage in sector i of an individual with endowment s is:

(3)
$$\ln w_i(s) = \ln \pi_i + \ln t_i(s)$$

The proportion of the population working in sector i is:

$$pr\left(i
ight)=\int_{\mathcal{L}_{i}}g\left(s
ight|\Theta
ight)ds, \hspace{0.2cm}i=1,2$$

Roy model assumes that $g(s | \Theta)$ and $t_i(s)$ are such that:

$$\begin{bmatrix} \ln t_1(s) \\ \ln t_2(s) \end{bmatrix} \quad N\left[\begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \Sigma \right]$$

Roy Model

Estimates ●000

Estimates of the Extended Roy Model

Non-manufacturing sector: predicted vs. observed log wage distribution



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Roy Model

Estimates ○●○○

Estimates of the Extended Roy Model

Manufacturing sector: predicted vs. observed log wage distribution



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Exploring the Importance of Aggregation Bias in Aggregate Wages

Simulation of a 1% increase in the energy price index

	Manufacturing Sector	Nonmanufacturing Sector	U.S. Aggregate
Year: 1972:			
 Percentage change in persons employed Percentage change in mean task or quality 	-1.854	1.320	
level for the employed population	.919	-1.496	
3. Percentage change in task price 4. Percentage change in observed average	- 1.480	.471	062*
wage (2 + 3)	561	- 1.025	950
Year: 1976:			
 Percentage change in persons employed Percentage change in mean task or quality 	-2.007	1.371	
level for the employed population	.886	-1.461	
 Percentage change in task price Percentage change in observed average 	- 1.480	.471	063*
wage (2 + 3)	594	990	939
Year: 1980:			
 Percentage change in persons employed Percentage change in mean task or quality 	- 1.993	1.244	
level for the employed population	953	-1 568	
3. Percentage change in task price 4. Percentage change in observed average	-1.480	.471	034*
wage (2 + 3)	527	997	949

NOTE.-The data sets on which the simulations are performed are defined in App. C.

* This is a weighted average of the task price change in each sector using the relative proportions employed in the sector in the year.

Assessing the Impact of Self Selection on Inequality in log Wages

Assessing the impact of self-selection on the means and variances of log wage rates for white males, 1980

	Prediction of Extended Roy Model	Actual 1980 Value	Random Assignment Economy Using 1980 Equilibrium Task Prices		
		Nonmanufacturing Sector	or		
Mean of log wages (M_1) Variance of log wages (σ_1) Proportion of population in sector (P_1)	1.054 .319 .619*	1.040 .323 .630	.651 .344 .619*		
		Manufacturing Sector			
Mean of log wages (M_2) Variance of log wages (σ_2) Proportion of population in sector (P_2)	1.199 .192 .200*	1.202 .201 .206	.968 .211 .200*		
		Economywide			
Mean of log wages $\left(\frac{P_1M_1 + P_2M_2}{P_1 + P_2}\right)$	1.089	1.079	.728		
Sum of within-sector variance $\left(\frac{P_1\sigma_1 + P_2\sigma_2}{P_1 + P_2}\right)$.288	.293	.311		
Between-sector variance $\left[\frac{P_1P_2(M_1 - M_2)^2}{(P_1 + P_2)^2}\right]$.003	.004	.018		
Total variance [†]	.291	.297	.329		

* The random assignment economy is restricted to have the proportion of people in each of the three sectors predicted by our model using 1980 equilibrium values. † Total variance = within-variance + between-variance

$$= \left(\begin{array}{c} \frac{P_1 \sigma_1 + P_2 \sigma_2}{P_1 + P_2} \end{array} \right) + \left[\begin{array}{c} \frac{P_1 P_2 (M_1 - M_2)^2}{(P_1 + P_2)^2} \end{array} \right].$$