

Growth Model with Automation and Endogenous Human Capital

Rong Fan

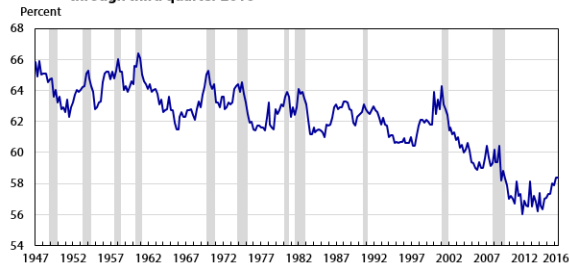
Indiana University

fanrong@iu.edu

May 7, 2022

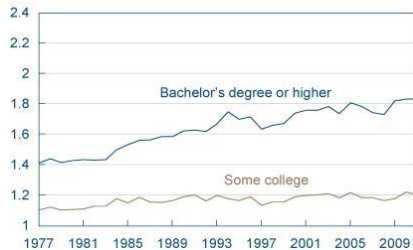
Motivation

Figure 1. Labor's share of output in the nonfarm business sector, first quarter 1947 through third quarter 2016



Source: Bureau of Labor Statistics

Figure 1. Wage Premiums for College



Source: Federal Reserve Bank of Cleveland

- Declining labor share
- Increasing college premium

Automation

- Generalized CES: Prettner and Strulikc (2020)
- Task model: Acemoglu and Autor (2011), Acemoglu and Restrepo (2018, 2019, 2020, 2021)

Education and technology

- Twin engines: Stokey (2018), Adao et al. (2020)
- Race: Goldin and Katz (2009), Grossman et al. (2020)

Empirical evidence

- AI Occupational Impact (AIOI) measured by Felten et al.(2019)
- Occupational Information Network (O*NET): bi-annually

$$y_{ijt} = \underbrace{t + t \times \text{AIOI}_i}_{\text{Time Trend}} + \underbrace{t \times i + t \times j + \alpha_i + \gamma_j}_{\text{Fixed effect}} + \epsilon_{ijt}$$

\downarrow
 Skill level j for occupation i

Skill	(1) Content	(2) Process	(3) Social	(4) Complex Problem
Time	-0.000415*** (0.000151)	-0.000441*** (0.000149)	-0.000411*** (0.000143)	0.000592*** (0.000200)
Time × AIOI	0.000464** (0.000228)	0.000543** (0.000225)	0.000485** (0.000216)	-0.000901*** (0.000303)
Fixed effect	Yes	Yes	Yes	Yes
Observations	174,720	116,480	174,720	29,120
R-squared	0.760	0.724	0.631	0.842

Standard errors in parentheses
 *** p<0.01, ** p<0.05, * p<0.1

- **How does automation affect skilled and unskilled workers?**

Direct effect on unskilled workers and ripple effect on skilled workers

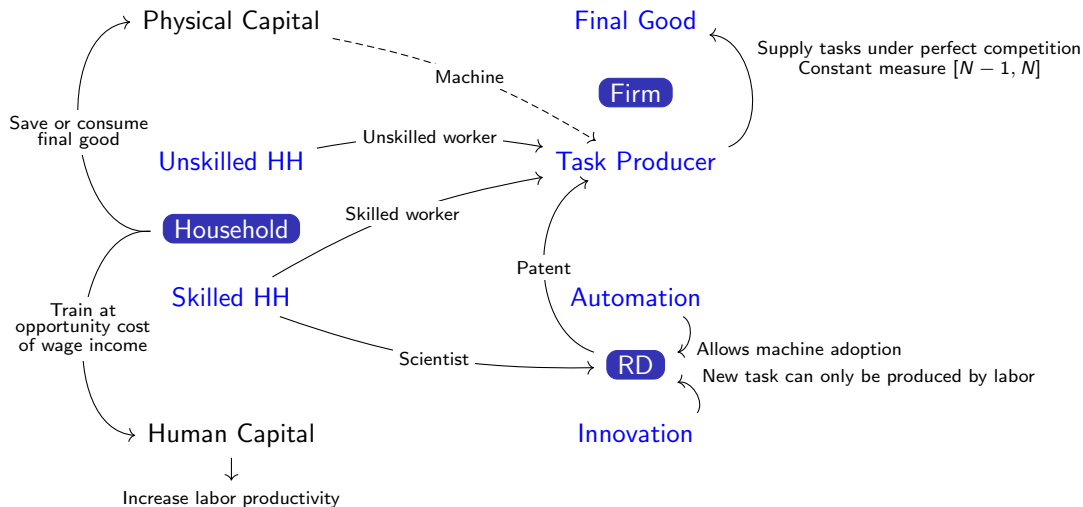
- **How do workers respond to automation?**

Uneven human capital investment

- **Is human capital important for understanding automation?**

Wage premium and inequality

Model



Human capital investment

Law of motion:

$$\dot{h}_j = \frac{(1 - l_j)^{\alpha_j}}{\mu_{hj}}, \quad j = \{H, L\}$$

↑ Training time

↓ Change in human capital

↓ Training cost

↓ Skilled or Unskilled

Human capital investment

Law of motion:

$$\dot{h}_j = \frac{(1 - l_j)^{\alpha_j}}{\mu_{hj}}, \quad j = \{H, L\}$$

Diagram illustrating the Law of Motion for Human Capital Investment:

- \dot{h}_j : Change in human capital
- $(1 - l_j)^{\alpha_j}$: Training time
- μ_{hj} : Training cost
- $j = \{H, L\}$: Skilled or Unskilled

- $\alpha_j \rightarrow 1$, learning or doing
- $\alpha_j \rightarrow 0$, learning by doing
- $\alpha_H > \alpha_L$, different learning ability

Human capital investment

Law of motion:

$$\dot{h}_j = \frac{(1 - l_j)^{\alpha_j}}{\mu_{hj}}, \quad j = \{H, L\}$$

Diagram illustrating the relationship between variables in the model:

- \dot{h}_j (Change in human capital) is related to $(1 - l_j)^{\alpha_j}$ (Training time) and μ_{hj} (Training cost).
- $j = \{H, L\}$ (Skilled or Unskilled) is related to μ_{hj} .

- $\alpha_j \rightarrow 1$, learning or doing
- $\alpha_j \rightarrow 0$, learning by doing
- $\alpha_H > \alpha_L$, different learning ability

Euler:
Trade off between
physical and
human capital

$$\underbrace{\frac{\delta \log \omega_j(h_j) l_j}{\delta h_j} \frac{\delta \dot{h}_j}{\delta (1-l_j)}}_{\text{Direct wage gain}} + \underbrace{\frac{\delta \log \omega_j(h_j)}{\delta h_j} \dot{h}_j + \frac{\delta \log \omega_j(h_j)}{\delta t}}_{\text{Return to human capital}} = r \quad \downarrow \text{Return to physical capital}$$

$$\frac{\delta \log \omega_j(h_j)}{\delta t} = g_{\omega|h}(g_N, g_I - g_N, g_{h-j})$$

\downarrow
 Growth
(+)

\downarrow
 Automation
(-)

\downarrow
 Human capital
(+)

Production factor allocation

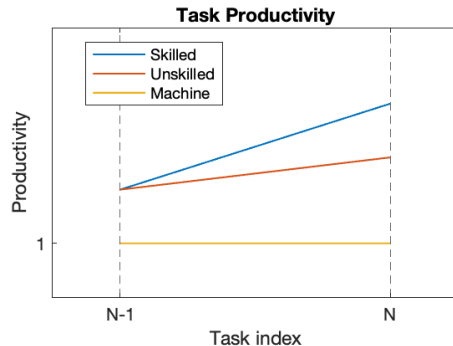
Factor productivity:

Machine: $\eta(i) = 1$

Skilled worker: $\gamma_H(i, h_H) = e^{B_H(i-1)+B_H(i-(N-1))} e^{b_H h_H}$

Unskilled worker: $\gamma_L(i, h_L) = e^{B_H(i-1)+B_L(i-(N-1))} e^{b_L h_L}$

\downarrow Task \downarrow Human capital



Production factor allocation

Factor productivity:

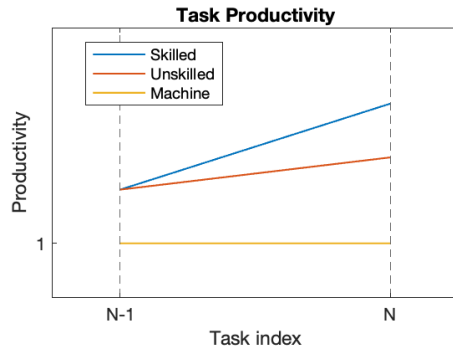
Machine: $\eta(i) = 1$

Skilled worker: $\gamma_H(i, h_H) = e^{B_H(i-1)+B_H(i-(N-1))} e^{b_H h_H}$

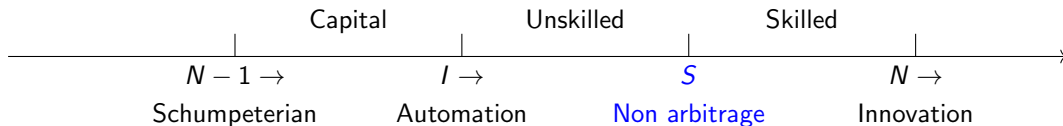
Unskilled worker: $\gamma_L(i, h_L) = e^{B_H(i-1)+B_L(i-(N-1))} e^{b_L h_L}$

↓
Task

↓
Human capital



Equilibrium allocation:



Directed research

Scientist productivity:

Cost = Return

$$\text{Innovation: } \dot{N} = \frac{1}{\mu_N} \epsilon_N^\lambda$$

$$\omega_H = \frac{\lambda}{\mu_N} \epsilon_N^{\lambda-1} P_N$$

$$\text{Automation: } \dot{I} = \frac{1}{\mu_I} \epsilon_I^\lambda, \lambda < 1$$

$$\omega_H = \frac{\lambda}{\mu_I} \epsilon_I^{\lambda-1} P_I$$

↓ ↓
Output Scientist

↓ ↓
High skill wage Patent value

Patent value:

$$\text{Innovation: } P_N = V_N(N) - V_I(N-1)$$

↓ ↓
Newest task profit Oldest task profit

$$\text{Automation: } P_I = V_I(I) - V_N(I)$$

↓ ↓
Automated task profit Unautomated task profit

Directed research

Scientist productivity:

$$\text{Innovation: } \dot{N} = \frac{1}{\mu_N} \epsilon_N^\lambda$$

$$\text{Automation: } \dot{I} = \frac{1}{\mu_I} \epsilon_I^\lambda, \lambda < 1$$

Output Scientist

Cost = Return

$$\omega_H = \frac{\lambda}{\mu_N} \epsilon_N^{\lambda-1} P_N$$

$$\omega_H = \frac{\lambda}{\mu_I} \epsilon_I^{\lambda-1} P_I$$

High skill wage Patent value

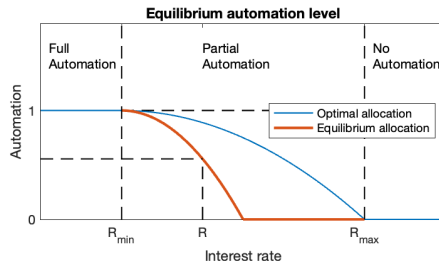
Patent value:

$$\text{Innovation: } P_N = V_N(N) - V_I(N-1)$$

Newest task profit Oldest task profit

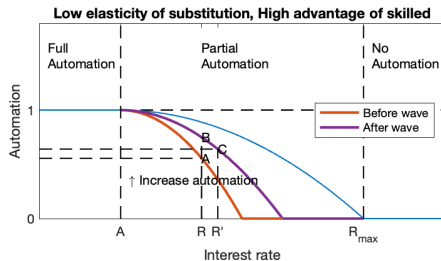
$$\text{Automation: } P_I = V_I(I) - V_N(I)$$

Automated task profit Unautomated task profit



- At optimal allocation, $P_I = 0$
- Equilibrium $<$ Optimal $\rightarrow P_I > 0$
- Equilibrium allocation is determined by non-arbitrage condition

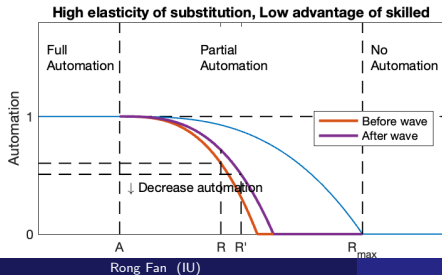
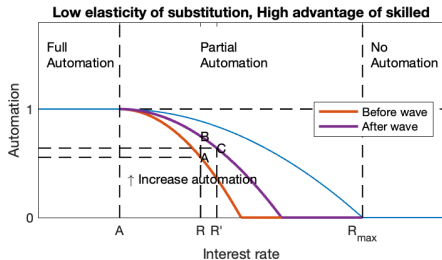
Technology wave



Lower automation cost: $\mu_I \downarrow$

- Productivity effect: $A \rightarrow B$
Automation \uparrow , Wage premium \uparrow
- Price effect: $B \rightarrow C$
Capital demand \uparrow , $R \uparrow$
Automation \downarrow , Wage premium \downarrow

Technology wave



Lower automation cost: $\mu_I \downarrow$

- Productivity effect: $A \rightarrow B$
Automation \uparrow , Wage premium \uparrow
- Price effect: $B \rightarrow C$
Capital demand \uparrow , $R \uparrow$
Automation \downarrow , Wage premium \downarrow

Strong price effect with strong ripple effect

- Unskilled workers can relocate more easily
- Higher elasticity of substitution between production factors
- Lower comparative advantage of skilled worker (productivity, human capital)

Calibration: external

Parameters calibrated externally

Parameter	Description	Value
θ	Intertemporal elasticity of substitution	0.8
δ	Depreciation rate	0.12
ϵ_H	High skill workers share	0.3
ϵ_L	Low skill workers share	0.7
μ_N	Innovation cost	1
μ_I	Automation cost	1
μ_h	Training cost	1

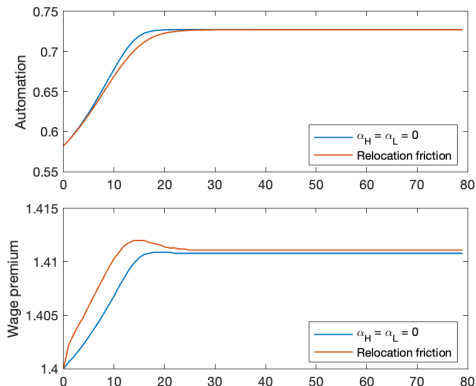
Calibration: internal

Parameters calibrated internally

Parameter	Description	Value	Targeted moment	Value
ρ	Discount rate	0.0196	Long run interest rate	0.04
η	Patent share	0.0684	RD and GDP ratio (1980)	0.0245
σ	Elasticity of substitution	3.112	RD growth rate	0.02
A	Capital productivity	0.1359	Short run interest rate	0.04
b_h	Human capital productivity	0.1885	Human capital growth rate	0.0055
B_H	High skill CA	0.6727	Wage premium (1980)	1.4
B_N	Low skill CA	0.4968	Labor share (1980)	0.64
λ	RD Decreasing return	0.7786	RD and GDP ratio (2000)	0.0265
z	Technology shock size	0.3472	Labor share (2000)	0.6

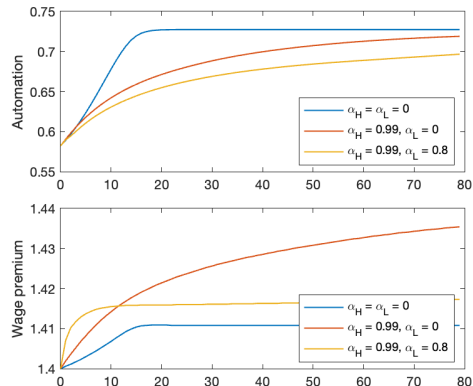
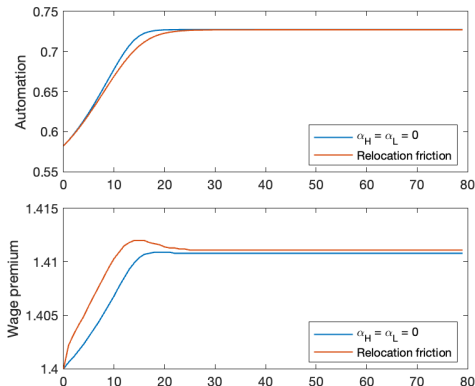
Data source: FRED

Transition: Automation and wage premium



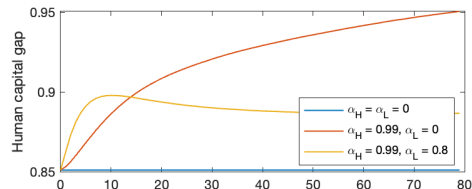
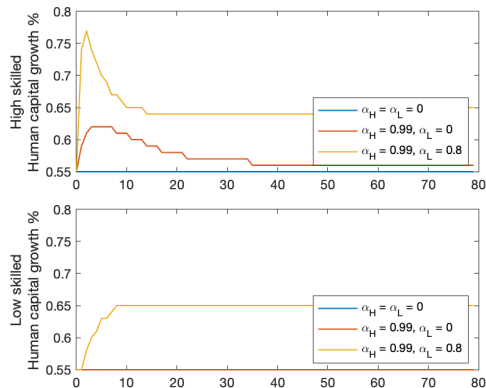
- Small wage premium change without human capital response.
- Limited improvement with relocation friction.

Transition: Automation and wage premium



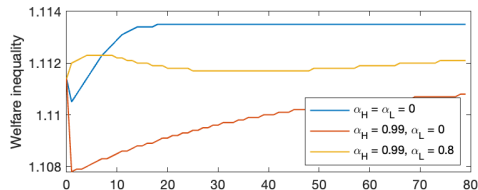
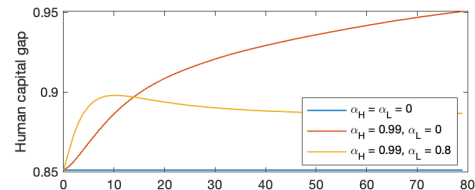
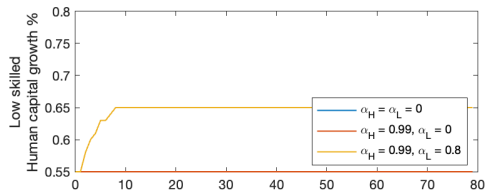
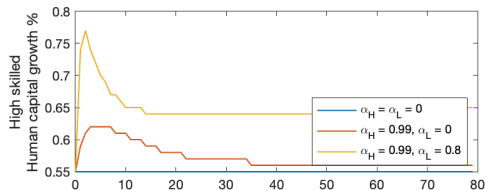
- Small wage premium change without human capital response.
- Limited improvement with relocation friction.
- Human capital investment lowers the automation level but raises the wage premium
- Change in wage premium depends on $\alpha_H - \alpha_L$.

Transition: Human capital



- Workers increase their human capital investment after the technology shock.
- Skilled worker respond more.

Transition: Human capital



- Workers increase their human capital investment after the technology shock.
- Skilled worker respond more.
- Human capital decreases the welfare inequality.

Main takeaways

- The uneven impacts of automation depend on the magnitude of ripple effect.
- Skilled workers respond more to a technology wave.
- Human capital investment explains the rise in the wage premium.
- Human capital investment decreases the welfare inequality.

Skilled HH problem

Skilled workers/Scientist: ϵ_H

$$\begin{aligned} & (\rho + (\theta - 1)g)V_{Ht}(k_H, h_H) - \frac{dV_{Ht}(k_H, h_H)}{dt} \\ & = \max_{c_H, l_H} u(c_H) + \frac{dV_{Ht}(k_H, h_H)}{dk_H} \dot{k}_H + \frac{dV_{Ht}(k_H, h_H)}{dh_H} \dot{h}_H \end{aligned}$$

Physical capital: $\dot{k}_H = (r - g)k_H + \omega_H L_H + \pi - c_H$

Human capital: $\dot{h}_H = \frac{(1 - l_H)^{\alpha_H}}{\mu_{hH}}, \quad 0 \leq \alpha_H \leq 1$

Labor supply: $\underbrace{L_H}_{\text{Workers}} = \epsilon_H(1 - l_H) - \underbrace{(\epsilon_N + \epsilon_I)}_{\text{Scientists}}$

Cost of human capital investment:

$$\mu_{hH} = f(h_{hH} - h_{hL})$$

back

Unskilled HH problem

Unskilled workers: ϵ_L

$$\begin{aligned} & (\rho + (\theta - 1)g)V_{Lt}(k_L, h_L) - \frac{dV_{Lt}(k_L, h_L)}{dt} \\ &= \max_{c_L, l_L} u(c_L) + \frac{dV_{Lt}(k_L, h_L)}{dk_L} \dot{k}_L + \frac{dV_{Lt}(k_L, h_L)}{dh_L} \dot{h}_L \end{aligned}$$

Physical capital: $\dot{k}_L = (r - g)k_L + \omega_L L_L - c_L$

Human capital: $\dot{h}_L = \frac{(1 - l_L)^{\alpha_L}}{\mu_{hL}}, \quad 0 \leq \alpha_L \leq \alpha_H$

Labor supply: $L_L = \epsilon_L(1 - l_L)$

Cost of human capital investment:

$$\mu_{hL} = \mu_h$$

back

Firm problem

Final good

$$\underbrace{Y}_{\text{Output}} = \underbrace{A}_{\text{TFP}} \left(\int_{N-1}^N \underbrace{y(i)^{\frac{\sigma-1}{\sigma}}}_{\text{task}} di \right)^{\frac{\sigma}{\sigma-1}}$$

Task: With automation constraint I

$$y(i) = \begin{cases} q(i)^\eta \left(k(i) + \gamma_L(i, h_L)l(i) + \gamma_H(i, h_H)h(i) \right)^{1-\eta}, & N-1 \leq i \leq I \\ \underbrace{q(i)^\eta}_{\text{Patent}} \underbrace{\left(\gamma_L(i, h_L)l(i) + \gamma_H(i, h_H)h(i) \right)^{1-\eta}}_{\text{Production factor}}, & I < i \leq N \end{cases}$$

back

Firm problem

Demand function

$$y(i) = A^{\sigma-1} Y p(i)^{\sigma}$$

Price = Cost of production

$$p(i) = \begin{cases} \phi \min\{R^{1-\eta}, \left(\frac{W_H}{\gamma_L(i, h_H)}\right)^{1-\eta}, \left(\frac{W_L}{\gamma_L(i, h_L)}\right)^{1-\eta}\}, & N-1 \leq i \leq I \\ \phi \min\{\left(\frac{W_H}{\gamma_L(i, h_H)}\right)^{1-\eta}, \left(\frac{W_L}{\gamma_L(i, h_L)}\right)^{1-\eta}\}, & I < i \leq N \end{cases}$$

where $\phi = \left(\frac{\psi}{\eta}\right)^{\eta} \left(\frac{1}{1-\eta}\right)^{1-\eta}$

back

Patent Value

$$\text{Innovation: } P_N(t) = V_N(N(t), t) - V_I(N(t) - 1, t)$$

$$\text{Automation: } P_I(t) = V_I(I(t), t) - V_N(I(t), t)$$

Present discounted value of future profit

$$\text{Task N using labor: } V_N(N, t) = \int_t^\infty e^{-\int_t^\tau r(s)ds} \pi(N, \tau) d\tau$$

$$\text{Task I using machine: } V_I(I, t) = \int_t^\infty e^{-\int_t^\tau r(s)ds} \pi(I, \tau) d\tau$$

Scientist productivity

$$\dot{N} = \frac{1}{\mu_N} \epsilon_N^\lambda \quad \dot{I} = \frac{1}{\mu_I} \epsilon_I^\lambda$$

back

Ripple effect

$$\underbrace{d\tilde{S}}_{\text{Ripple effect}} = \underbrace{\frac{1}{(\sigma-1)\Lambda(\tilde{S})}(b_L dh_L - b_H dh_H)}_{\text{Human capital}} + \underbrace{\frac{1}{\Lambda(\tilde{S})}\delta_L(\tilde{I})d\tilde{I}}_{\text{Automation}}$$
$$\Lambda(\tilde{S}) = \underbrace{\left(\frac{d \ln \Gamma_L}{d\tilde{S}} - \frac{d \ln \Gamma_H}{d\tilde{S}}\right)}_{\text{Non arbitrage}} + \underbrace{\frac{\sigma}{\sigma-1}(B_H - B_L)}_{\text{ES and CA}}$$
$$\underbrace{d \log \omega}_{\text{Wage premium}} = (B_H - B_L)d\tilde{S}$$

back