

Directed technical change with capital-embodied technologies

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Overview

- 1. Capital-embodied technical change: what and why?
- 2. A model of directed technical change with embodiment
- 3. Optimal policies in the calibrated model
- 4. Conclusions and recommendations



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Capital-embodied technical change: what and why?





Technical change and economic growth

Neo-classical aggregate production function Y = F(L, K)

- Positive but decreasing marginal returns to each factor input
- If constant returns to scale: $\lambda F(L, K) = F(\lambda L, \lambda K)$

Neo-classical growth model $Y_t = A_t F(L_t, K_t)$

Considering A_t only as technology (ignore institutions, etc.):

- A_t is a non-rival input
- *A_t* may be *non-excludable*
 - Typically *partially* excludable with use of patents, secrecy, etc.



1.2.

Capital-embodiment of technologies

R&D mostly directed at new or improved *products* esp. capital *equipment*

- Good evidence for declining real equipment prices
- US productivity growth >60% capital-embodied (Greenwood *et al.*)
- Macro literature focuses on IT revolution
- But clear relevance to new and old energy technologies
 - Gas turbines, solar panels, wind turbines, LED bulbs, batteries, ...



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Why is capital-embodiment important?

Diffusion of new technologies requires investments

Models with disembodied TC ignore this dependence

User cost of capital increases with the innovation rate

- Return on real assets must cover
 - Required return on equity
 - Physical depreciation
 - Expected change in asset price
- TC causes *declining* asset prices <> obsolescence costs
 - => If rates of TC varies between sectors or over time, so should rates of economic depreciation



Models with *directed* technical change (DTC)

Single sector with factor-augmenting TC

$$Y = F\left(A_{L}L, A_{K}K\right)$$

- where F(.) is not Cobb-Douglas and A_L and A_K are disembodied technologies
- TFP growth in heterogeneous sectors

$$Y_{t} = \prod_{i} Y_{i,t}^{\alpha_{i}} = \prod_{i} \left[A_{i,t} F_{i} \left(L_{i,t}, K_{i,t} \right) \right]^{\alpha_{i}}$$

• Technical change can be partially embodied, depending on the nature of *K*. We will return to this at the end.



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Other forms of embodiment

Embodiment in workers or firms

- Learning to use new equipment
- Incremental ('engineering') improvements or adaptations of existing technologies

Arguably, bounded by invented technologies (Young, 1993)





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Modelling capital-embodied environmentally directed technical change



Directed technical change with embodiment

Framework of Acemoglu et al., 2012 (AABH)

Clean, dirty and final production

$$Y_{_t} = \left(Y_{_{c,t}}^{(arepsilon-1)/arepsilon} + Y_{_{d,t}}^{(arepsilon-1)/arepsilon}
ight)^{arepsilon/(arepsilon-1)}, \quad arepsilon > 1$$

$$Y_{_{j,t}} = L_{_{j,i,t}}^{_{1-lpha}} \int_{_{0}}^{^{1}} A_{_{j,i,t}}^{_{1-lpha}} x_{_{j,i,t}}^{^{lpha}} di, \quad j \in \left\{c,d
ight\}$$

- Profit-driven R&D to improve clean or dirty intermediates x_{i,i,t}
- Emissions from dirty sector -> climate -> damage costs

Embodying technical change:

- Clean and dirty *capital* goods: $Y_{j,t} = L_{j,i,t}^{1-\alpha} \int_0^1 k_{j,i,t}^{\alpha} di \quad j \in \{c,d\}$
- Technical change
 "investment specific" (Krusell, 1998): $k_{j,i,t} = (1 \delta)k_{j,i,t-1} + A_{j,i,t}z_{j,i,t}$



Embodiment and obsolescence costs

Rental rate per unit of effective capital of type (j,i)

$$r_{j,i,t} \approx \left(\delta + i_t + g_{j,i,t}\right) / \left(\alpha A_{j,i,t}\right), \quad g_{j,i,t} \equiv A_{j,i,t+1} / A_{j,i,t} - 1$$

- $1/A_{j,i,t}$ cost per unit of effective capital
- $1/\alpha$ monopolists' mark-up over investment costs
- $g_{j,i,t}$ growth rate of technology

Response of clean to dirty output ratio to a step change in $g_{c,t}$

$$\frac{Y_{\scriptscriptstyle c,t}}{Y_{\scriptscriptstyle d,t}} \approx \left(1+\tau_{\scriptscriptstyle t}\right)^{\varepsilon} \left(\frac{i_{\scriptscriptstyle t}+\delta+g_{\scriptscriptstyle c,t}}{i_{\scriptscriptstyle t}+\delta+g_{\scriptscriptstyle d,t}}\right)^{-\alpha\varepsilon} \left(\frac{A_{\scriptscriptstyle c,t}}{A_{\scriptscriptstyle d,t}}\right)^{\varepsilon}$$

- Decreases with increase in $g_{c,t}$ once-off short-run effect
- Increases with growth of $A_{c,t}$ dominant long run effect



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2.2

Research and development

Research and development firms

 One R&D firm per capital good. Hires scientists to improve technology building on previous sector-average technology

• Knowledge frontier as in AABH:
$$A_{j,i,t} = (1 + \eta_j s_{j,i,t}) A_{j,t-1}$$

Symmetry

- Deterministic progress implies symmetry of firms within each sector:
- Complete spillovers and deterministic progress unrealistic, but convenient
 - Concerned with productivity differences between not within sectors.

Spillovers

- Knowledge spillovers between sectors empirically significant but not primarily between clean and dirty energy technologies
- => Assume spillovers from an exogenously growing technology frontier

$$A_{j,t} = \left(1 + \eta_j \phi \left(\frac{A_{t-1}^{exogenous} - A_{j,t-1}}{A_{t-1}^{exogenous}}\right)^{\varphi} s_{j,t}\right) A_{j,t-1}$$



Scientists are the sole input to R&D

 Fixed supply of scientists, equally capable of working on any technology

Profit-maximising allocation of scientists

- R&D firms seek to maximise their profits
 - Capture PV of investment in their technology in the current period
 - Do not capture future value because of inter-temporal spillovers
- Profits depend only on level of raw investment not on the level of output as in AABH: $\pi_{j,t} = z_{j,t} (s_{j,t}) (1-\alpha) / \alpha$

Hiring more scientists in sector *j* improves *j* technologies

- Increases demand for *effective* capital $k_{i,t}$ and hence $A_{i,t}z_{j,t}$
- Decreases *raw* capital $z_{i,t}$ per unit of effective capital



Optimal policies in the calibrated model



Structure of optimal policies

Capital rental subsidy corrects monopoly distortion

- Optimal subsidy rate = α (inverse of the mark-up factor)
 - Could use (time-varying) investment subsidies with equivalent economic effect

Dirty tax corrects emissions externality

- Marginal cost of a unit increase in CO₂ concentration
- Less present value of future CO₂ removals (by biogeophysical sinks)

R&D subsidy internalises intertemporal tech spillovers

- Fixed R&D supply implies subsidy can be phased out once clean technology is sufficiently advanced that clean profits exceed dirty
- Intersectoral spillovers make R&D in backward sector relatively more productive => subsidy rate need to induce clean R&D is lower



Optimal policies: effects of embodiment & spillovers

Policies induce immediate switch to clean R&D in all models



Dirty tax rates

- Similar initial rates but rising faster
 Including spillovers
- Lower initial rates but rising faster because faster clean progress lowers aggregate costs



R&D subsidy rates

- Higher rates & slower phase-out
 Including spillovers
- Reduces required subsidies



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3.2

Embodiment & spillovers: temperature & consumption

0%

-70%



Atmospheric temperature

- Mitigation more costly => Significantly higher peak temperature
 Including spillovers
- Aggregate mitigation costs decline faster
 => Temperature peaks earlier & lower

Consumption

0

····· DISEMB

- EMB

50

EMB & EX SPILL

 Consumption losses reduced in first century but increased in second
 Including spillovers

100

Years

150

200

250

 Consumption losses smaller and decline in second century



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Embodiment & spillovers: output & investment





Dirty output

Jump in clean capital rents vs. dirty
 => initial fall (rise) in clean (dirty) output
 => persistent lag in mitigation

Including spillovers

- Initial response unchanged
- Dirty output declines faster thereafter

Investment

Jump in clean capital rents vs. dirty
 => initial fall (rise) in clean (dirty) investment

Including spillovers

Faster growth of clean technology
 => accelerated demand for clean capital in long run



Conclusions and recommendations

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Key findings

Capital-embodiment can substantially alter dynamic responses:

- Diffusion of new technologies requires investments
- Technical progress generates obsolescence costs
- Returns to R&D depend on investment not output

Increasing the rate of clean TC relative to dirty

- Naturally, beneficial in the long run
- Perverse level effect in the short(er) run

Optimal mitigation timing

Investment & R&D decisions intimately linked



Adding a third, non-energy-intensive sector

- Additional margin of substitution
- Realistic composition effects => plausible macroeconomic costs
- Endogenous intersectoral spillovers

Two region or small open economy model

- New technologies embodied in imported equipment
- Disembodied international knowledge spillovers in R&D



Embodied technologies \Leftrightarrow heterogeneous capital

- Rarely considered in CGE models, although likely widely relevant
 - May be explained in significant part by data limitations
- Considered in some bottom-up energy (sub-)models
 - But linked to learning curves, not R&D-driven technical change

Embodiment distinct from irreversibility

 Irreversibility of investment binds only for "large" shocks to "narrowly defined" industries (or capital asset classes)



Future work: embodying technologies in ICES

Region- and sector-specific rates of TFP growth

- Exogenous rates (for now)
 - Based mainly on EU KLEMS database

Introduce heterogeneous capital

- Structures
- Several classes of equipment
 - Based on EU KLEMS &/or US BEA capital flows
- Distinguish "green" (wind turbines, PV modules, etc.)

Obsolescence costs

 Dependence of regional demand for investment on rate of change in real investment prices



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